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ABSTRACT

Medical Cannabis Availability and Mental Health: Evidence from New York's Medical Cannabis Program^{*}

Evidence on cannabis legalization's effects on mental health remains scarce, despite both rapid increases in cannabis use and an ongoing mental health crisis in the United States. We use granular geographic data to estimate medical cannabis dispensary availability's effects on self-reported mental health in New York state from 2011 through 2021 using a two-stage difference-in-differences approach to minimize bias introduced from the staggered opening of dispensaries. Our findings rule out that medical cannabis availability had negative effects on mental health for the adult population overall. We also find that medical cannabis availability reduced past-month self-reported poor mental health days by nearly 10%—3.37 percentage points—among adults 65 and above. These results suggest medical cannabis access has positive health impacts for older populations, likely through pain relief.

JEL Classification:H75, I12, I18, I31, K32, K42Keywords:cannabis, mental health

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^{*} New York BRFSS data used in this analysis may be requested from the New York State Department of Health at https://www.health.ny.gov/statistics/brfss/. We thank the New York Department of Health for sharing those data. We have provided dispensary data as a supplement to this article. Drake and Nagy were supported by the National Institute on Drug Abuse (NIDA) under award K01DA051761. We are grateful to Dr. Catherine Maclean for her helpful feedback.

1. Introduction

How consumers respond to differences in firm location is a classic topic in economics (Hotelling 1929; Debreu 1959; Starrett 1974). A substantial literature examines how consumers' distance from health care firms influences the firm from which they receive medical care (McGuire, Kreif, and Smith 2021; Grabowski et al. 2013; McClellan, McNeil, and Newhouse 1994) and how location affects downstream health outcomes (Avdic, Lundborg, and Vikström 2024; Bertoli and Grembi 2017; Kessler and McClellan 2000; Lu and Slusky 2016; Slusky 2017; Lu and Slusky 2019; Ellison et al. 2021; Card, Fenizia, and Silver 2023). Shorter distances between consumers and products lower the non-monetary costs of products to consumers, which increase use and then has a greater potential impact on downstream outcomes. Understanding these downstream effects is of particular importance as newer therapeutics with mixed evidence of efficacy become more widely available to consumers.

One such case of particular importance is the distribution of legal cannabis through licensed dispensaries and its effects on mental health. Cannabis use continues to surge as nearly 40 states have passed medical cannabis laws (ProCon.org 2024). An estimated 62 million Americans—22% of the population aged 12 and older—used cannabis in 2022, nearly doubling from 11% in 2010. Cannabis use decreases with age, ranging from 37% among persons aged 18 to 34 to 8% among persons aged 65 and over in 2022, though recent increases in use have occurred across age groups (Substance Abuse and Mental Health Services Administration 2023).

The clinical literature suggests cannabis use may exacerbate depressive and mood disorders (National Academies of Sciences, Engineering, and Medicine 2017). However, cannabis is often used and promoted as a wellness product to treat mental health conditions. Roughly 50% and 34% of medical cannabis consumers reported using it to treat and anxiety and depression (Kosiba, Maisto, and Ditre 2019). That rapid increases in cannabis use are occurring amidst a rapidly worsening mental health crisis exacerbated by the COVID-19 pandemic (McGinty et al. 2020; Mehra et al. 2023) increases the need for evidence on how cannabis availability affects mental health.

In this paper, we examine how mental health is affected by medical cannabis dispensary availability in the context of New York's medical cannabis program. Our analysis takes place prior to recreational cannabis legalization. Nearly all states with medical cannabis laws require consumers to purchase medical cannabis at licensed medical cannabis dispensaries (ProCon.org

2024), limiting access to consumers with the ability to travel to dispensaries. Consumers' access therefore does not just depend just on the state in which they live, but also where they live in that state.

We have two key findings. First, we can rule out that medical cannabis availability causes meaningful increases in self-reported poor mental health among the adult population overall. Second, we find that medical cannabis availability has meaningful and robust positive effects on self-reported mental health among persons aged 65 and above, who experience 3.48 percentage point decrease in having self-reported past-month poor mental health days, a roughly 10% decrease from a baseline of 36.3%. While these findings suggest positive mental health effects of medical cannabis availability, the positive effects are concentrated among the portion of the age distribution least likely to use cannabis. Roughly 2.29% of New Yorkers aged 65 and above reported using cannabis in 2015—the year before New York opened its first medical dispensaries. About twice as many, 4.49%, did so in 2019 (Substance Abuse and Mental Health Services Administration 2023). High stigma surrounding cannabis use among this population, however, may lead to underestimates of older adults' cannabis use (Dahlke et al. 2024), which helps to explain the relatively large magnitude of our findings.

This study integrates two streams of emerging cannabis economics literature by using refined geographic measures to examine cannabis availability's effects on mental health. Three prior studies (Borbely et al. 2022; Sabia, Swigert, and Young 2017; Nicholas and Maclean 2019), to our knowledge, have examined cannabis legalization's effects on mental health, though they rely on state variation in cannabis laws to identify cannabis availability's effects rather than substate variation in dispensary variation, as we do. Other recent studies have used sub-state geographic variation in dispensary availability to examine other health outcomes (Wang et al. 2022; Smith 2020; Conyers and Ayres 2020; Ambrose, Cowan, and Rosenman 2021).

2. Background

The federal government criminalized the possession and sale of cannabis with the Marihuana Tax Act of 1937 (Dufton 2017). California was the first state to legalize medical cannabis in 1996. Few other states followed until the 2009 Ogden Memorandum, which deprioritized federal enforcement of cannabis law. As of May 2024, 38 states have legalized medical cannabis nearly three million Americans possessed licenses to use medical cannabis by 2020 (Boehnke et al. 2022)—and 24 states have legalized recreational cannabis (ProCon.org 2024). Medical

cannabis only increases cannabis use once a state opens dispensaries (Hollingsworth, Wing, and Bradford 2022).

New York implemented its medical cannabis program in 2016. New Yorkers over age 21 with a qualifying medical condition had to obtain a license to purchase medical cannabis. Unlike some other states, New York did not allow at-home cultivation of medical cannabis. New York's qualifying medical conditions were typical of medical cannabis programs, including chronic pain and post-traumatic stress disorder but notably excluding anxiety (ProCon.org 2024). The program enrolled over 100,000 persons by 2019 and over 150,000 by 2021 (New York State Department of Health 2022; 2019).

New York legalized recreational cannabis on March 31, 2021. The sale of recreational cannabis became legal in 2022. Unlicensed recreational cannabis dispensaries proliferated throughout New York when recreational cannabis sales became legal in 2022, making it impossible to establish a spatial relationship between cannabis availability and health outcomes (Southall 2024). For these reasons, we end our study period in 2021. Cannabis delivery services, another potential threat to our study design, did not become available until 2023 (Southall 2023).

3. Methods

3.1. Data

Our primary data are the 2011-2021 New York Behavioral Risk Factor Surveillance System (BRFSS) (New York Department of Health 2024). The BRFSS is a nationally coordinated, stateadministered telephone survey that collects data on health, health-related risk behaviors, and respondents' sociodemographic characteristics. Importantly for this study, the New York BRFSS data report respondents home ZIP codes, which are not available in the national BRFSS data. We include all BRFSS respondents with non-missing ZIP codes and responses to mental health-related questions.

We also use data on listings of New York medical cannabis dispensaries obtained from the New York Department of Health (New York Office of Cannabis Management 2024). These data list all medical cannabis dispensaries operating in New York, but they do not list the dates those dispensaries opened. We conducted an extensive review of local media sources to identify dispensaries' opening dates. We provide the New York Department of Health data with opening dates and sources in the supplementary appendix. Importantly for our study design, we found no

evidence that any medical cannabis dispensaries closed during the study period, though they did sometimes change ownership.

We used Microsoft's BingMaps application programming interface to convert BRFSS respondents' five-digit ZIP codes and dispensaries' addresses to longitude-latitude coordinates (Microsoft 2024). BingMaps identifies the most central point in each ZIP code that is located on a road. We then calculated drive times between each ZIP code and dispensary, again using BingMaps. These drive time calculations consider the built environment, including road structure, speed limits, stop signs and traffic lights, and congestion. We calculate drive times at 9AM on a Wednesday to simulate a standard commute.

3.2. Measures

Our primary outcomes examine self-reported mental health. The BRFSS surveys mental health by asking, "Now thinking about your mental health, which includes stress, depression, and problems with emotions, for how many days during the past 30 days was your mental health not good?" (New York Department of Health 2024). We use these questions to construct three measures of mental health, per prior studies (Slabaugh et al. 2017; Cree et al. 2020; Taylor 2000; Liu et al. 2018): (1) an indicator that identifies whether respondents reported any past-month poor mental health days; (2) the number of past-month poor mental health days; and (3) an indicator for whether respondents reported *frequent mental distress*, which corresponds to 14 or more days of poor mental health.

Our treatment is an indicator for whether respondents had an open medical cannabis dispensary within a 30-minute drive by the start of the quarter-year in which they were surveyed. This measure identifies whether medical cannabis was available within a standard commute time (U.S. Census Bureau 2021). We examine other drive time thresholds in sensitivity analyses. Prior research found that medical cannabis laws only increase cannabis laws when they are accompanied with open medical dispensaries (Hollingsworth, Wing, and Bradford 2022), indicating that our classification of ZIP codes near dispensaries as "treated" and far away ZIP codes as "comparisons" is valid.

3.3. Empirical Strategy

While we employ a two-way fixed effects (TWFE) approach, our preferred specification is a two-stage difference-in-differences (2SDID) approach because of the staggered implementation of medical cannabis dispensary openings across New York.

We estimate self-reported mental health for respondent i in ZIP code z in quarter-year t in our TWFE specification such that

$$MH_{izt} = \beta D_{zt} + \theta_z + \tau_t + \epsilon_{izt}, \qquad (1)$$

where D_{zt} is an indicator for a medical dispensary being open within a 30-minute drive of ZIP code z in year t; θ_z and τ_t are ZIP code and quarter-year fixed effects; and ϵ_{izt} is a countyclustered error term. We cluster standard errors at the county level because of likely spatial correlation among adjoining ZIP codes affected by dispensary openings (Slusky 2017; Lu and Slusky 2019; 2016). We also include a vector of respondent-level sociodemographic characteristics included in the BRFSS in some specifications.

A Bacon decomposition (Goodman-Bacon 2021) shows only 6% of the two-by-two comparisons comprising our overall TWFE estimate are problematic, indicating that an approach robust to staggered implementation is necessary.

In the first stage of the 2SDID approach (Gardner 2022; Powell 2022), we estimate our outcome for all untreated observations, from which we create residualized fixed effects. In the second stage, we estimate our outcome using the whole sample with the residualized fixed effects from the first stage. We estimate this model using a generalized method of moments approach such that:

$$MH_{izt}|(D=0) = \theta_z + \tau_t + \mu_{izt} \tag{1}$$

$$MH_{izt} = \beta D_{zt} + \hat{\theta}_z + \hat{\tau}_t + \omega_{izt}.$$
 (2)

Our treatment coefficient β is identified under the parallel trends assumption. Here, that assumption requires that BRFSS self-reported mental health responses in ZIP codes with 30-minute medical cannabis dispensary availability would have evolved similarly to ZIP codes without dispensary availability in a counterfactual state where they had not been treated. The 2SDID approach does not require additional identifying assumptions, though it does explicitly use the parallel trends assumption in its use of residualized fixed effects from untreated observations. We lend support to the parallel trends assumption and examine dynamics in treatment effects by estimating event study analogs of our TWFE and 2SDID approaches.

4. Results

4.1. Descriptive Statistics

Our sample consists of 180,539 respondent-quarter-years in 1,982 ZIP codes in all 62 New York counties from 2011 to 2021, after eliminating 20,133 (10.0%) observations with missing mental (4,764) health outcomes and/or ZIP codes (15,806). The BRFSS oversamples older adults: 33.3% of our sample is aged 65 and above. Approximately 33.7% of the sample reported some pastmonth poor mental health days in 2015, the year prior to treatment, with a mean of 3.52 days and 10.8% reporting frequent mental distress. National Survey on Drug Use and Health data indicate that 15% of adult New Yorkers used cannabis in 2015—the year before medical cannabis implementation—and increased to 17% in 2019 (Substance Abuse and Mental Health Services Administration 2023).

Appendix Figure 1 shows there were not large differences in outcomes between treatment and comparison groups in the 12 quarter-years prior to treatment. Appendix Table 1 reports differences in outcomes and sociodemographic characteristics in 2015, the year preceding treatment. Differences by age and sex are relatively small in magnitude. Those by race and ethnicity are not, suggesting sub-analyses by race and ethnicity may be inappropriate.

4.2. Baseline Results

Table 1 shows results from TWFE and 2SDID models with and without covariates. Having any self-reported past-month poor mental health days is negatively associated with 30-minute medical cannabis dispensary availability in all specifications. While the magnitudes of estimates are large (-1.54 to -4.92 percentage points (pp)), they are only significant for the 2SDID specification without covariates, which estimates a 2.31 (95% CI = -4.25 to -0.37) percentage point decrease ("intent-to-treat") in having any poor mental health days. Pre-trends tests support the parallel trends assumption. While we prefer 2SDID to TWFE, we are hesitant to conclude that medical cannabis availability decreases poor mental health days because of the smaller magnitude and lack of significance in the 2SDID results with covariates. However, given the 2SDID with covariates 95% confidence interval, -3.36 to 0.28, we can rule out that medical cannabis availability causes meaningful increases in having past-month self-reported mental health days. We estimate negative but noisy associations between medical cannabis availability and number of days of cannabis use and \geq 14 days of past-month cannabis use.

4.3. Stratified Results

We proceed by stratifying our results by age, sex, and race and ethnicity for our 2SDID specifications with and without covariates in Table 2 and Appendix Table 3, respectively. We find

that medical cannabis availability affects a 3.37pp (95% CI = -6.09 to -0.65) decrease in any poor mental health days among respondents aged 65 and above. Results are similar with covariates. We do not observe other significant associations between medical cannabis availability and mental health for other outcomes or other demographic groups. Most estimates are negative but noisy. Appendix Table 3 reports pre-trends tests for all demographic subgroups, which are not significant in most cases, including respondents aged 65 and above.

4.4. Robustness Checks

We vary the drive time availability threshold in Appendix Table 4. Results are consistent at 25, 30, and 35 minutes. We also exclude Manhattan from our sample in Appendix Table 5 because Manhattan residents use public transit more than the rest of the state. Our results are consistent in magnitude to our original findings, though the 2SDID estimate for the whole sample without covariates is no longer statistically significant, further suggesting this finding should be interpreted cautiously.

5. Discussion

In this study, we evaluate the geographic availability of medical cannabis' impact on selfreported mental health in New York state. We have two key findings. First, we can rule out that medical cannabis availability had meaningful negative effects on self-reported mental health among the adult population overall. This finding is broadly consistent with prior literature (Borbely et al. 2022; Sabia, Swigert, and Young 2017). However, some prior studies have found negative mental health effects for younger adults (Wang et al. 2022; Borbely et al. 2022). Our findings do not rule out such negative effects; rather, our estimates for younger adults are imprecise because younger adults are underrepresented in the BRFSS.

Our second key finding is that medical cannabis availability affected a 3.48 percentage point decrease in persons aged 65 and above reporting having any past-month poor mental health days, a nearly 10% decrease from a baseline of 36.3%. We did not find other significant effects on other measures of self-reported mental health or among different demographic subgroups. Collectively, these results suggest medical cannabis availability has limited mental health effects on the population at large, with considerable mental health benefits for older adults.

Pain relief is the likely mechanism through which medical cannabis availability reduces poor mental health days among older adults. Cannabis is effective for treating chronic neuropathic pain—the most common reported reason for medical cannabis use, the most common qualifying condition for medical cannabis licenses, and an especially prevalent condition among older adults (National Academies of Sciences, Engineering, and Medicine 2017; Nunberg et al. 2011; Boehnke et al. 2022; Nahin 2015). Prior studies are consistent with a pain relief mechanism. They have found that: (1) cannabis legalization reduces opioid prescribing and use among older adults, suggesting cannabis may increasingly be used as a substitute for pain relief (Bao et al. 2023; Bradford et al. 2018; Bradford and Bradford 2018); and (2) cannabis legalization reduces pain and increases work capacity among older adults (Sabia, Swigert, and Young 2017; Nicholas and Maclean 2019; Abouk et al. 2023).

Our finding that self-reported poor mental health days decrease by 3.37pp among persons aged 65 and above is large. Cannabis use among persons 65 and above did roughly double from the year before the state implemented medical cannabis to four years after, from 2.29% in 2015 to 4.49% in 2019 (Substance Abuse and Mental Health Services Administration 2023). Yet this increase is insufficient to accommodate such a relatively large effect—how might such an increase be possible? One explanation is that cannabis use is underreported among older persons due to stigma surrounding cannabis use, which is strongest among older populations (Dahlke et al. 2024). Another explanation is selection—older adults that stand to the most from medical cannabis tend to enroll in medical cannabis programs. There also may be intra-family spillover effects: People with a partner that benefits from cannabis use may experience improved mental health due to the partner's improved wellbeing, and/or begin using cannabis themselves but not report it because they do not have a medical cannabis license. The confidence interval of our estimate also leaves room for considerably smaller effects. We note our finding is consistent with prior literature finding improved mental health among older adults after cannabis legalization (Nicholas and Maclean 2019; Sabia, Swigert, and Young 2017; Borbely et al. 2022).

Our results demonstrate that cannabis firms' locations have a large effect on whether consumers experience downstream effects of introducing medical cannabis into a market. As states continue to regulate the locations of cannabis dispensaries, they should consider how nearby populations will be affected by cannabis availability. Such considerations are especially important given several states' efforts to advance health equity by supporting opening dispensaries in minority communities (Kilmer 2019).

References

- Abouk, Rahi, Keshar M. Ghimire, Johanna Catherine Maclean, and David Powell. 2023. "Pain Management and Work Capacity: Evidence From Workers' Compensation and Marijuana Legalization." *Journal of Policy Analysis and Management* 42 (3): 737–70. https://doi.org/10.1002/pam.22479.
- Ambrose, Christopher A., Benjamin W. Cowan, and Robert E. Rosenman. 2021. "Geographical Access to Recreational Marijuana." *Contemporary Economic Policy* 39 (4): 778–807. https://doi.org/10.1111/coep.12518.
- Avdic, Daniel, Petter Lundborg, and Johan Vikström. 2024. "Does Health Care Consolidation Harm Patients? Evidence from Maternity Ward Closures." *American Economic Journal: Economic Policy* 16 (1): 160–89. https://doi.org/10.1257/pol.20200734.
- Bao, Yuhua, Hao Zhang, Eduardo Bruera, Russell Portenoy, William E. Rosa, M. Carrington Reid, and Hefei Wen. 2023. "Medical Marijuana Legalization and Opioid- and Pain-Related Outcomes Among Patients Newly Diagnosed With Cancer Receiving Anticancer Treatment." JAMA Oncology 9 (2): 206–14. https://doi.org/10.1001/jamaoncol.2022.5623.
- Bertoli, Paola, and Veronica Grembi. 2017. "The Life-saving Effect of Hospital Proximity." *Health Economics* 26 (52): 78–91.
- Boehnke, Kevin F., Owen Dean, Rebecca L. Haffajee, and Avinash Hosanagar. 2022. "U.S. Trends in Registration for Medical Cannabis and Reasons for Use From 2016 to 2020: An Observational Study." *Annals of Internal Medicine* 175 (7): 945–51. https://doi.org/10.7326/M22-0217.
- Borbely, Daniel, Otto Lenhart, Jonathan Norris, and Agnese Romiti. 2022. "Marijuana Legalization and Mental Health." Working Paper 15729. IZA Discussion Papers. https://www.econstor.eu/handle/10419/267466.
- Bradford, Ashley C., and W. David Bradford. 2018. "The Impact of Medical Cannabis Legalization on Prescription Medication Use and Costs under Medicare Part D." *The Journal of Law and Economics* 61 (3): 461–87. https://doi.org/10.1086/699620.
- Bradford, Ashley C., W. David Bradford, Amanda Abraham, and Grace Bagwell Adams. 2018. "Association Between US State Medical Cannabis Laws and Opioid Prescribing in the Medicare Part D Population." *JAMA Internal Medicine* 178 (5): 667–72. https://doi.org/10.1001/jamainternmed.2018.0266.
- Card, David, Alessandra Fenizia, and David Silver. 2023. "The Health Impacts of Hospital Delivery Practices." *American Economic Journal: Economic Policy* 15 (2): 42–81. https://doi.org/10.1257/pol.20210034.
- Conyers, Gregory, and Ian Ayres. 2020. "A Lottery Test of the Effect of Dispensaries on Emergency Room Visits in Arizona." *Health Economics* 29 (8): 854–64. https://doi.org/10.1002/hec.4013.
- Cree, Robyn A., Catherine A. Okoro, Matthew M. Zack, and Eric Carbone. 2020. "Frequent Mental Distress Among Adults, by Disability Status, Disability Type, and Selected Characteristics — United States, 2018." *Morbidity and Mortality Weekly Report* 69 (36): 1238–43. https://doi.org/10.15585/mmwr.mm6936a2.
- Dahlke, Sherry, Jeffrey I. Butler, Kathleen F. Hunter, Madeline Toubiana, Maya R. Kalogirou, Shovana Shrestha, Rashmi Devkota, Joanna Law, and Melissa Scheuerman. 2024. "The Effects of Stigma: Older Persons and Medicinal Cannabis." *Qualitative Health Research*, February, 10497323241227419. https://doi.org/10.1177/10497323241227419.

Debreu, Gerard. 1959. "Topological Methods in Cardinal Utility Theory."

- Dufton, Emily. 2017. Grass Roots: The Rise and Fall and Rise of Marijuana in America. Basic Books.
- Ellison, Jacqueline, Kevin Griffith, Madalyn Thursby, David J. G. Slusky, and Jacob Bor. 2021. "The Impact of Driving Time to Family Planning Facilities on Preventive Service Use in Ohio." *American Journal of Preventive Medicine* 60 (4): 542–45. https://doi.org/10.1016/j.amepre.2020.11.009.
- Gardner, John. 2022. "Two-Stage Differences in Differences." arXiv. https://doi.org/10.48550/arXiv.2207.05943.
- Goodman-Bacon, Andrew. 2021. "Difference-in-Differences with Variation in Treatment Timing." *Journal of Econometrics*, Themed Issue: Treatment Effect 1, 225 (2): 254–77. https://doi.org/10.1016/j.jeconom.2021.03.014.
- Grabowski, David C., Zhanlian Feng, Richard Hirth, Momotazur Rahman, and Vincent Mor. 2013. "Effect of Nursing Home Ownership on the Quality of Post-Acute Care: An Instrumental Variables Approach." *Journal of Health Economics* 32 (1): 12–21. https://doi.org/10.1016/j.jhealeco.2012.08.007.
- Hollingsworth, Alex, Coady Wing, and Ashley C. Bradford. 2022. "Comparative Effects of Recreational and Medical Marijuana Laws on Drug Use among Adults and Adolescents." *The Journal of Law* and Economics 65 (3): 515–54. https://doi.org/10.1086/721267.
- Hotelling, Harold. 1929. "Stability in Competition." *The Economic Journal* 39 (153): 41–57. https://doi.org/10.2307/2224214.
- Kessler, Daniel P., and Mark B. McClellan. 2000. "Is Hospital Competition Socially Wasteful?" *The Quarterly Journal of Economics* 115 (2): 577–615.
- Kilmer, Beau. 2019. "How Will Cannabis Legalization Affect Health, Safety, and Social Equity Outcomes? It Largely Depends on the 14 Ps." *The American Journal of Drug and Alcohol Abuse* 45 (6): 664–72. https://doi.org/10.1080/00952990.2019.1611841.
- Kosiba, Jesse D., Stephen A. Maisto, and Joseph W. Ditre. 2019. "Patient-Reported Use of Medical Cannabis for Pain, Anxiety, and Depression Symptoms: Systematic Review and Meta-Analysis." *Social Science & Medicine* 233 (July):181–92. https://doi.org/10.1016/j.socscimed.2019.06.005.
- Liu, Jin, Ning Jiang, Amy Z. Fan, and Rebecca Weissman. 2018. "Alternatives in Assessing Mental Healthcare Disparities Using the Behavioral Risk Factor Surveillance System." *Health Equity* 2 (1): 199–206. https://doi.org/10.1089/heq.2017.0056.
- Lu, Yao, and David J. G. Slusky. 2016. "The Impact of Women's Health Clinic Closures on Preventive Care." American Economic Journal: Applied Economics 8 (3): 100–124. https://doi.org/10.1257/app.20140405.
- 2019. "The Impact of Women's Health Clinic Closures on Fertility." American Journal of Health Economics 5 (3): 334–59. https://doi.org/10.1162/ajhe_a_00123.
- McClellan, Mark, Barbara J. McNeil, and Joseph P. Newhouse. 1994. "Does More Intensive Treatment of Acute Myocardial Infarction in the Elderly Reduce Mortality?: Analysis Using Instrumental Variables." *JAMA* 272 (11): 859–66. https://doi.org/10.1001/jama.1994.03520110039026.
- McGinty, Emma E., Rachel Presskreischer, Kelly E. Anderson, Hahrie Han, and Colleen L. Barry. 2020. "Psychological Distress and COVID-19–Related Stressors Reported in a Longitudinal Cohort of US Adults in April and July 2020." JAMA 324 (24): 2555–57. https://doi.org/10.1001/jama.2020.21231.

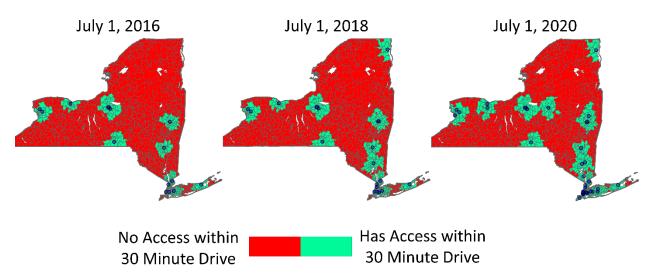
- McGuire, Finn, Noemi Kreif, and Peter C. Smith. 2021. "The Effect of Distance on Maternal Institutional Delivery Choice: Evidence from Malawi." *Health Economics* 30 (9): 2144–67. https://doi.org/10.1002/hec.4368.
- Mehra, Kamna, Jennifer Rup, Jessica L. Wiese, Tara Marie Watson, Sarah Bonato, and Sergio Rueda. 2023. "Changes in Self-Reported Cannabis Use during the COVID-19 Pandemic: A Scoping Review." *BMC Public Health* 23 (November):2139. https://doi.org/10.1186/s12889-023-17068-7.
- Microsoft. 2024. "Bing Maps Dev Center Bing Maps Dev Center." 2024. https://www.bingmapsportal.com/.
- Nahin, Richard L. 2015. "Estimates of Pain Prevalence and Severity in Adults: United States, 2012." *The Journal of Pain* 16 (8): 769–80. https://doi.org/10.1016/j.jpain.2015.05.002.
- National Academies of Sciences, Engineering, and Medicine. 2017. *The Health Effects of Cannabis and Cannabinoids: The Current State of Evidence and Recommendations for Research*. Washington, D.C.: National Academies Press. https://doi.org/10.17226/24625.
- New York Department of Health. 2024. "Behavioral Risk Factor Surveillance System (BRFSS)." 2024. https://www.health.ny.gov/statistics/brfss/.
- New York Office of Cannabis Management. 2024. "Dispensing Facilities." 2024. https://cannabis.ny.gov/dispensing-facilities.
- New York State Department of Health. 2019. "New York State Medical Marijuana Program." December 18, 2019.

https://web.archive.org/web/20191218203739/https://www.health.ny.gov/regulations/medical_mar ijuana/.

- . 2022. "New York State Medical Marijuana Program." January 7, 2022. https://web.archive.org/web/20220107225626/https://www.health.ny.gov/regulations/medical_mar ijuana/.
- Nicholas, Lauren Hersch, and Johanna Catherine Maclean. 2019. "The Effect of Medical Marijuana Laws on the Health and Labor Supply of Older Adults: Evidence from the Health and Retirement Study." *Journal of Policy Analysis and Management* 38 (2): 455–80. https://doi.org/10.1002/pam.22122.
- Nunberg, Helen, Beau Kilmer, Rosalie Liccardo Pacula, and James R. Burgdorf. 2011. "An Analysis of Applicants Presenting to a Medical Marijuana Specialty Practice in California." *Journal of Drug Policy Analysis* 4 (1). https://doi.org/10.2202/1941-2851.1017.
- Powell, David. 2022. "The Labor Supply Consequences of the Opioid Crisis." SSRN, June. https://doi.org/10.2139/ssrn.3899329.
- ProCon.org. 2024. "State-by-State Medical Marijuana Laws." Britannica. 2024. https://medicalmarijuana.procon.org/legal-medical-marijuana-states-and-dc/.
- Sabia, Joseph J., Jeffrey Swigert, and Timothy Young. 2017. "The Effect of Medical Marijuana Laws on Body Weight." *Health Economics* 26 (1): 6–34. https://doi.org/10.1002/hec.3267.
- Slabaugh, S. Lane, Mona Shah, Matthew Zack, Laura Happe, Tristan Cordier, Eric Havens, Evan Davidson, Michael Miao, Todd Prewitt, and Haomiao Jia. 2017. "Leveraging Health-Related Quality of Life in Population Health Management: The Case for Healthy Days." *Population Health Management* 20 (1): 13–22. https://doi.org/10.1089/pop.2015.0162.
- Slusky, David J. G. 2017. "Defunding Women's Health Clinics Exacerbates Hispanic Disparity in Preventive Care." *Economics Letters* 156 (July):61–64. https://doi.org/10.1016/j.econlet.2017.04.013.

- Smith, Rhet A. 2020. "The Effects of Medical Marijuana Dispensaries on Adverse Opioid Outcomes." *Economic Inquiry* 58 (2): 569–88. https://doi.org/10.1111/ecin.12825.
- Southall, Ashley. 2023. "Expecting Cannabis Boom, New York Lays Down the Rules." *The New York Times*, September 13, 2023, sec. New York. https://www.nytimes.com/2023/09/13/nyregion/cannabis-marijuana-regulations-ny.html.
 - ———. 2024. "New York's Marijuana Scorecard: 85 Legal Shops, 2,000 Illegal Ones." *The New York Times*, March 22, 2024. https://www.nytimes.com/2024/03/22/nyregion/nyc-cannabis-dispensary-legal.html.
- Starrett, David A. 1974. "Principles of Optimal Location in a Large Homogeneous Area." *Journal of Economic Theory* 9 (4): 418–48. https://doi.org/10.1016/0022-0531(74)90044-1.
- Substance Abuse and Mental Health Services Administration. 2023. "HHS, SAMHSA Release 2022 National Survey on Drug Use and Health Data." Text. Hhs.Gov. November 13, 2023. https://www.hhs.gov/about/news/2023/11/13/hhs-samhsa-release-2022-national-survey-drug-use-health-data.html.
- Taylor, Virginia Ross. 2000. "Measuring Healthy Days : Population Assessment of Health-Related Quality of Life." *Centers for Disease Control*. https://stacks.cdc.gov/view/cdc/6406.
- U.S. Census Bureau. 2021. "Travel Time to Work in the United States: 2019." Census.Gov. March 18, 2021. https://www.census.gov/library/publications/2021/acs/acs-47.html.
- Wang, George Sam, Christine Buttorff, Asa Wilks, Daniel Schwam, Gregory Tung, and Rosalie Liccardo Pacula. 2022. "Impact of Cannabis Legalization on Healthcare Utilization for Psychosis and Schizophrenia in Colorado." *International Journal of Drug Policy* 104 (June):103685. https://doi.org/10.1016/j.drugpo.2022.103685.

Figure 1. 30-Minute Medical Cannabis Dispensary Availability in New York State on July 1st in 2016, 2018, and 2020



Notes: Blue dots represent medical cannabis dispensaries. Car-based drive times are calculated from the geographic center of each ZIP code using Microsoft's Bing Maps application programming interface, which accounts for road structure, speed limits, traffic lights, stop signs, congestion, etc.

Outcome or	Model (N = 180,539)			
Model Property	TWFE, No	TWFE with	2SDID, No	2SDID with
	Covariates (1)	Covariates (2)	Covariates (3)	Covariates (4)
Self-Reported Mer	ntal Poor Mental Ou	utcomes		
Any (0/1)	-4.92 (2.62)	-4.04 (2.77)	-2.31* (0.99)	-1.54 (0.93)
Count (#)	-0.56 (0.39)	-0.48 (0.4)	-0.20 (0.16)	-0.11 (0.15)
≥2 Weeks (%)	-1.77 (1.6)	-1.58 (1.65)	-0.36 (0.62)	-0.09 (0.61)
F-Tests of Pre-Tre	nds			
Any (0/1)	-3.16 (2.69)	-3.07 (2.87)	-0.02 (0.03)	-0.02 (0.03)
Count (#)	-0.40 (0.41)	-0.41 (0.41)	0.00 (0.00)	0.00 (0.00)
≥2 Weeks (%)	-1.48 (1.68)	-1.52 (1.71)	-0.01 (0.02)	-0.01 (0.02)
Model Properties				
Covariates		Х		Х
Specification	TWFE	TWFE	Gardner	Gardner
Fixed Effects	ZIP code,	ZIP code,	ZIP code,	ZIP code,
	quarter-year	quarter-year	quarter-year	quarter-year
Clustering	County	County	County	County

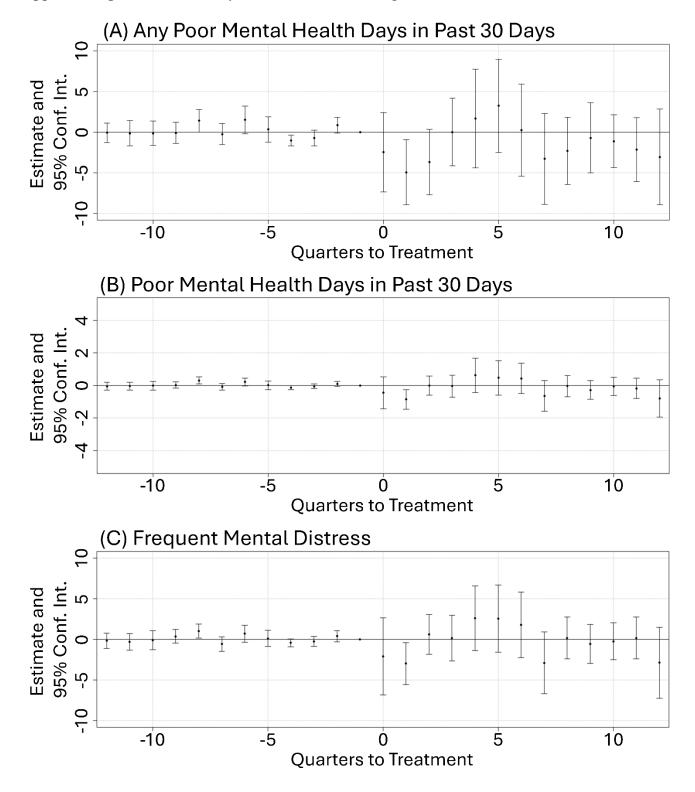
Table 2. Baseline Regression Results

Notes: Two-way fixed effects (TWFE) models are estimated using ordinary least squares; twostage differences-in-differences is estimated per Gardner (2021). We use survey weights in all models and cluster standard errors at the county level. See methods for a description of covariates.

* p < 0.05 ** p < 0.01 *** p < 0.001

A. Models without Covariates Age Group 18-24 -1.15 (4.24) -0.82 (0.65) -2.36 (3.04 25-34 -1.77 (3.04) 0.20 (0.56) 1.12 (2.28) 35-44 -0.60 (2.79) 0.13 (0.44) 1.14 (1.78) 45-54 -2.35 (2.12) -0.09 (0.36) -0.32 (1.4) 55-64 0.34 (1.77) -0.10 (0.34) -0.66 (1.28) 65+ -3.37* (1.39) -0.39 (0.23) -0.79 (0.9) Sex Female -2.91* (1.29) -0.39 (0.22) -0.05 (0.89) Male -2.1 (1.42) -0.05 (0.22) 0.05 (0.89) Male -2.1 (1.42) -0.05 (0.22) 0.05 (0.89) Non-Hispanic White 0.27 (1.07) 0.13 (0.17) 0.61 (0.68) Non-Hispanic Black 0.44 (5.92) -0.23 (1.29) 1.04 (5.5) Other -6.53 (3.44) -0.58 (0.68) 0.23 (2.76) B. Models with Covariates Age Group 18-24 -0.40 (4.30) -0.69 (0.63) -2.13 (2.98 25-34 -1.33 (2.88) 0.18 (0.52) 0.99 (2.16) 35-64 0.77 (1.67) 0.05 (0.32) <	Subgroup	Self-Reported Poor Mental Health Days in PAny DaysNumber of Days			
Age Group $18-24$ -1.15 (4.24) -0.82 (0.65) -2.36 (3.04) $25-34$ -1.77 (3.04) 0.20 (0.56) 1.12 (2.28) $35-44$ -0.60 (2.79) 0.13 (0.44) 1.14 (1.78) $45-54$ -2.35 (2.12) -0.09 (0.36) -0.32 (1.4) $55-64$ 0.34 (1.77) -0.10 (0.34) -0.66 (1.28) $65+$ -3.37* (1.39) -0.39 (0.23) -0.79 (0.9) Sex Female -2.91* (1.29) -0.39 (0.23) -0.94 (0.91) Male -2.1 (1.42) -0.05 (0.22) 0.05 (0.89) Race & Ethnicity Non-Hispanic White 0.27 (1.07) 0.13 (0.17) 0.61 (0.68) Non-Hispanic Black 0.44 (5.92) -0.23 (1.29) 1.04 (5.5) Other -6.53 (3.44) -0.58 (0.68) 0.23 (2.76) B. Models with Covariates Age Group 18-24 -0.40 (4.30) -0.69 (0.63) -2.13 (2.98 $25-34$ -1.33 (2.88) 0.18 (0.52) 0.99 (2.16) 35-44 -1.26 (2.76) -0.10 (0.43) 0.45 (1.77) $35-64$ 0.77 (1.67) 0.05 (0.32) -0.04 (1.24) 65+	Subgroup	Any Days	Number of Days	≥2 Weeks	
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$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Age Group				
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	18-24	-1.15 (4.24)	-0.82 (0.65)	-2.36 (3.04)	
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	25-34	-1.77 (3.04)	0.20 (0.56)	1.12 (2.28)	
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	35-44	-0.60 (2.79)	0.13 (0.44)	1.14 (1.78)	
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Age Group $18-24$ $-0.40 (4.30)$ $-0.69 (0.63)$ $-2.13 (2.98$ $25-34$ $-1.33 (2.88)$ $0.18 (0.52)$ $0.99 (2.16)$ $35-44$ $-1.26 (2.76)$ $-0.10 (0.43)$ $0.45 (1.77)$ $45-54$ $-2.17 (2.03)$ $-0.07 (0.34)$ $-0.46 (1.33)$ $55-64$ $0.77 (1.67)$ $0.05 (0.32)$ $-0.04 (1.24)$ $65+$ $-3.48* (1.38)$ $-0.40 (0.23)$ $-0.90 (0.90)$ Sex Female $-2.06 (1.19)$ $-0.31 (0.21)$ $-0.72 (0.87)$ Male $-1.18 (1.4)$ $0.07 (0.22)$ $0.42 (0.89)$ Race & Ethnicity Non-Hispanic White $0.57 (0.99)$ $0.18 (0.16)$ $0.74 (0.65)$ $0.72 (5.06)$	Other	-6.53 (3.44)	-0.58 (0.68)	0.23 (2.76)	
18-24 $-0.40 (4.30)$ $-0.69 (0.63)$ $-2.13 (2.98)$ $25-34$ $-1.33 (2.88)$ $0.18 (0.52)$ $0.99 (2.16)$ $35-44$ $-1.26 (2.76)$ $-0.10 (0.43)$ $0.45 (1.77)$ $45-54$ $-2.17 (2.03)$ $-0.07 (0.34)$ $-0.46 (1.33)$ $55-64$ $0.77 (1.67)$ $0.05 (0.32)$ $-0.04 (1.24)$ $65+$ $-3.48* (1.38)$ $-0.40 (0.23)$ $-0.90 (0.90)$ SexFemale $-2.06 (1.19)$ $-0.31 (0.21)$ $-0.72 (0.87)$ Male $-1.18 (1.4)$ $0.07 (0.22)$ $0.42 (0.89)$ Race & EthnicityNon-Hispanic White $0.57 (0.99)$ $0.18 (0.16)$ $0.74 (0.65)$ Non-Hispanic Black $1.36 (5.92)$ $-0.25 (1.17)$ $0.72 (5.06)$		B. Models	with Covariates		
25-34 $-1.33(2.88)$ $0.18(0.52)$ $0.99(2.16)$ $35-44$ $-1.26(2.76)$ $-0.10(0.43)$ $0.45(1.77)$ $45-54$ $-2.17(2.03)$ $-0.07(0.34)$ $-0.46(1.33)$ $55-64$ $0.77(1.67)$ $0.05(0.32)$ $-0.04(1.24)$ $65+$ $-3.48*(1.38)$ $-0.40(0.23)$ $-0.90(0.90)$ SexFemale $-2.06(1.19)$ $-0.31(0.21)$ $-0.72(0.87)$ Male $-1.18(1.4)$ $0.07(0.22)$ $0.42(0.89)$ Race & EthnicityNon-Hispanic White $0.57(0.99)$ $0.18(0.16)$ $0.74(0.65)$ Non-Hispanic Black $1.36(5.92)$ $-0.25(1.17)$ $0.72(5.06)$	Age Group				
35-44 $-1.26(2.76)$ $-0.10(0.43)$ $0.45(1.77)$ $45-54$ $-2.17(2.03)$ $-0.07(0.34)$ $-0.46(1.33)$ $55-64$ $0.77(1.67)$ $0.05(0.32)$ $-0.04(1.24)$ $65+$ $-3.48*(1.38)$ $-0.40(0.23)$ $-0.90(0.90)$ Sex FemaleFemale $-2.06(1.19)$ $-0.31(0.21)$ $-0.72(0.87)$ Male $-1.18(1.4)$ $0.07(0.22)$ 0.42 (0.89)Race & Ethnicity Non-Hispanic White $0.57(0.99)$ $0.18(0.16)$ $0.74(0.65)$ Non-Hispanic Black $1.36(5.92)$ $-0.25(1.17)$ $0.72(5.06)$	18-24	-0.40 (4.30)	-0.69 (0.63)	-2.13 (2.98)	
45-54 $-2.17 (2.03)$ $-0.07 (0.34)$ $-0.46 (1.33)$ $55-64$ $0.77 (1.67)$ $0.05 (0.32)$ $-0.04 (1.24)$ $65+$ $-3.48* (1.38)$ $-0.40 (0.23)$ $-0.90 (0.90)$ Sex Female $-2.06 (1.19)$ $-0.31 (0.21)$ $-0.72 (0.87)$ Male $-1.18 (1.4)$ $0.07 (0.22)$ $0.42 (0.89)$ Race & Ethnicity Non-Hispanic White $0.57 (0.99)$ $0.18 (0.16)$ $0.74 (0.65)$ Non-Hispanic Black $1.36 (5.92)$ $-0.25 (1.17)$ $0.72 (5.06)$	25-34	-1.33 (2.88)	0.18 (0.52)	0.99 (2.16)	
55-64 $0.77 (1.67)$ $0.05 (0.32)$ $-0.04 (1.24)$ $65+$ $-3.48* (1.38)$ $-0.40 (0.23)$ $-0.90 (0.90)$ Sex Female $-2.06 (1.19)$ $-0.31 (0.21)$ $-0.72 (0.87)$ Male $-1.18 (1.4)$ $0.07 (0.22)$ $0.42 (0.89)$ Race & Ethnicity Non-Hispanic White $0.57 (0.99)$ $0.18 (0.16)$ $0.74 (0.65)$ Non-Hispanic Black $1.36 (5.92)$ $-0.25 (1.17)$ $0.72 (5.06)$	35-44	-1.26 (2.76)	-0.10 (0.43)	0.45 (1.77)	
55-64 $0.77 (1.67)$ $0.05 (0.32)$ $-0.04 (1.24)$ $65+$ $-3.48* (1.38)$ $-0.40 (0.23)$ $-0.90 (0.90)$ Sex Female $-2.06 (1.19)$ $-0.31 (0.21)$ $-0.72 (0.87)$ Male $-1.18 (1.4)$ $0.07 (0.22)$ $0.42 (0.89)$ Race & Ethnicity Non-Hispanic White $0.57 (0.99)$ $0.18 (0.16)$ $0.74 (0.65)$ Non-Hispanic Black $1.36 (5.92)$ $-0.25 (1.17)$ $0.72 (5.06)$	45-54	-2.17 (2.03)	-0.07 (0.34)	-0.46 (1.33)	
Sex Female -2.06 (1.19) -0.31 (0.21) -0.72 (0.87) Male -1.18 (1.4) 0.07 (0.22) 0.42 (0.89) Race & Ethnicity Non-Hispanic White 0.57 (0.99) 0.18 (0.16) 0.74 (0.65) Non-Hispanic Black 1.36 (5.92) -0.25 (1.17) 0.72 (5.06)	55-64	0.77 (1.67)	0.05 (0.32)	-0.04 (1.24)	
Female -2.06 (1.19) -0.31 (0.21) -0.72 (0.87) Male -1.18 (1.4) 0.07 (0.22) 0.42 (0.89) Race & Ethnicity 0.07 (0.22) 0.42 (0.65) Non-Hispanic White 0.57 (0.99) 0.18 (0.16) 0.74 (0.65) Non-Hispanic Black 1.36 (5.92) -0.25 (1.17) 0.72 (5.06)	65+	-3.48* (1.38)	-0.40 (0.23)	-0.90 (0.90)	
Male -1.18 (1.4) 0.07 (0.22) 0.42 (0.89) Race & Ethnicity 0.07 (0.22) 0.42 (0.89) Non-Hispanic White 0.57 (0.99) 0.18 (0.16) 0.74 (0.65) Non-Hispanic Black 1.36 (5.92) -0.25 (1.17) 0.72 (5.06)	Sex				
Race & Ethnicity 0.18 (0.16) 0.74 (0.65) Non-Hispanic Black 1.36 (5.92) -0.25 (1.17) 0.72 (5.06)	Female	-2.06 (1.19)	-0.31 (0.21)	-0.72 (0.87)	
Non-Hispanic White0.57 (0.99)0.18 (0.16)0.74 (0.65)Non-Hispanic Black1.36 (5.92)-0.25 (1.17)0.72 (5.06)	Male	-1.18 (1.4)	0.07 (0.22)	0.42 (0.89)	
Non-Hispanic Black 1.36 (5.92) -0.25 (1.17) 0.72 (5.06)	Race & Ethnicity				
1	Non-Hispanic White	0.57 (0.99)	0.18 (0.16)	0.74 (0.65)	
Other -5.04 (3.42) -0.64 (0.67) -0.12 (2.71	Non-Hispanic Black	1.36 (5.92)	-0.25 (1.17)	0.72 (5.06)	
p < 0.05 ** p < 0.01 *** p < 0.001			-0.64 (0.67)	-0.12 (2.71)	

Table 3. Stratified Subgroup Analyses for 2SDID Model



Appendix Figure 1. Event Study of Baseline Gardner Regression

	Pre-Treatment Mean (2015Q1-2015Q4) Mean (SD)					
	Treated by 2018Q3	Not Treated by 2018Q3	Difference			
Characteristic	(N = 741)	(N = 691)	(SE)			
Self-Reported Poor Mental Health Days Within the Past 30 Days						
Any (%)	33.2	36.2	2.98 (.01)			
Count (#)	3.70	3.70	0.00 (.24)			
≥ 2 Weeks (%)	11.6	11.5	-0.09 (.01)			
Sociodemographic Charac	eteristics (%)					
Age						
18-24	44.3	45.6	1.30 (.04)			
25-34	37.3	40.9	3.61 (.03)			
35-44	36.7	38.5	1.86 (.02)			
45-54	36.0	36.4	0.43 (.02)			
55-64	31.0	32.9	1.92 (.02)			
65+	21.6	26.3	4.70 (.01)			
Female	38.0	39.2	1.28 (.01)			
Race & Ethnicity						
Non-Hispanic White	32.9	36.7	3.74 (.01)			
Non-Hispanic Black	22.3	34.2	12.00 (.04)			
Other	42.0	36.5	-5.46 (.03)			
Has Children	34.7	37.1	2.43 (.02)			
Unemployed	52.0	51.0	-0.97 (.02)			
Student	53.6	47.6	-6.01 (.05)			
Retired	22.2	26.3	4.06 (.01)			
Income≥\$50K	27.4	33.8	6.44 (.01)			
≥ Some College	33.1	35.7	2.59 (.01)			
Married	28.0	30.3	2.27 (.01)			
Uninsured	27.6	37.4	9.86 (.03)			

Appendix Table 1. Comparison of Sample Means between Treatment and Comparison Groups

Notes: Sample sizes refer to unique ZIP codes. Respondent-level differences are calculated using bivariate linear regressions with county-clustered error terms.

	Model	
	TWFE	2SDID
Dispensary Availability within	30 Minutes	
Treated	-4.04 (2.77)	-3.85 (7.04)
Leads (F-Test)	-3.07 (2.87)	-2.80 (2.93)
Sociodemographic Characterist	ics	
Age		
18-24		
25-34	-2.74* (1.09)	-1.65 (1.57)
35-44	-6.71*** (1.15)	-4.18* (1.68)
45-54	-8.60*** (1.13)	-6.77*** (1.59)
55-64	-13.19*** (1.15)	-10.97*** (1.58)
65+	-20.33*** (1.24)	-18.41*** (1.68)
Female	8.01*** (0.44)	8.08*** (0.60)
Race & Ethnicity		
Non-Hispanic White		
Non-Hispanic Black	-6.53*** (0.81)	-5.47*** (1.34)
Other	-4.55*** (0.69)	-2.64** (0.98)
Has Children	-0.77 (0.53)	0.13 (0.82)
Unemployed	16.15*** (0.70)	17.78*** (1.03)
Student	5.15*** (1.31)	6.26** (1.93)
Retired	-1.75** (0.62)	-0.60 (0.86)
Income \geq \$50K	-0.30 (0.50)	-1.87* (0.73)
≥ Some College	1.45** (0.49)	0.16 (0.71)
Married	-7.79*** (0.49)	-7.20*** (0.68)
Uninsured	-2.31** (0.86)	-1.29 (1.22)

Appendix Table 2. Full Results for TWFE, Gardner Models with Covariates

* p < 0.05 ** p < 0.01 *** p < 0.001

Subgroup	Any Days	Number of Days	ast 30 Days Outco >2 Weeks	
	1 2 4 9 5			
	A. With	out Covariates		
Age Group				
18-24	-1.15 (4.24)	-0.82 (0.65)	-2.36 (3.04)	
25-34	-1.77 (3.04)	0.20 (0.56)	1.12 (2.28)	
35-44	-0.60 (2.79)	0.13 (0.44)	1.14 (1.78)	
45-54	-2.35 (2.12)	-0.09 (0.36)	-0.32 (1.40)	
55-64	0.34 (1.77)	-0.10 (0.34)	-0.66 (1.28	
65+	-3.37* (1.39)	-0.39 (0.23)	-0.79 (0.90)	
Sex				
Female	-2.91* (1.29)	-0.39 (0.23)	-0.94 (0.91)	
Male	-2.10 (1.42)	-0.05 (0.22)	0.05 (0.89)	
Race & Ethnicity				
Non-Hispanic White	0.27 (1.07)	0.13 (0.17)	0.61 (0.68)	
Non-Hispanic Black	0.44 (5.92)	-0.23 (1.29)	1.04 (5.50)	
Other	-6.53 (3.44)	-0.58 (0.68)	0.23 (2.76)	
	B. Wit	h Covariates		
Age Group				
18-24	-0.40 (4.30)	-0.69 (0.63)	-2.13 (2.98	
25-34	-1.33 (2.88)	0.18 (0.52)	0.99 (2.16)	
35-44	-1.26 (2.76)	-0.10 (0.43)	0.45 (1.77)	
45-54	-2.17 (2.03)	-0.07 (0.34)	-0.46 (1.33	
55-64	0.77 (1.67)	0.05 (0.32)	-0.04 (1.24	
65+	-3.48* (1.38)	-0.40 (0.23)	-0.90 (0.90)	
Sex				
Female	-2.06 (1.19)	-0.31 (0.21)	-0.72 (0.87	
Male	-1.18 (1.40)	0.07 (0.22)	0.42 (0.89)	
Race & Ethnicity				
Non-Hispanic White	0.57 (0.99)	0.18 (0.16)	0.74 (0.65)	
Non-Hispanic Black	1.36 (5.92)	-0.25 (1.17)	0.72 (5.06)	
Other	-5.04 (3.42)	-0.64 (0.67)	-0.12 (2.71	

Appendix Table 3. F-Tests of Stratified Subgroup Analyses for 2SDID Model

* p < 0.05 ** p < 0.01 *** p < 0.001

Model		
(2)	(3)	
-2.31* (0.99)	-2.16* (1.06)	
-0.20 (0.16)	-0.24 (0.17)	
-0.36 (0.62)	-0.59 (0.66)	
-0.02 (0.03)	-0.02 (0.02)	
0.00 (0.00)	0.00 (0.00)	
-0.01 (0.02)	-0.01 (0.01)	
30	35	

Appendix Table 4. Results with Different Drive Time Thresholds for 2SDID Model without Covariates

Outcome or	Model (N = 180,539)			
Model Property	TWFE, No	TWFE with	2SDID, No	2SDID with
	Covariates (1)	Covariates (2)	Covariates (3)	Covariates (4)
Self-Reported Mer	ntal Poor Mental Ou	utcomes		
Any (0/1)	-3.92 (2.81)	2.53* (1.00)	-1.80 (0.93)	-3.92 (2.81)
Count (#)	-0.44 (0.41)	-0.19 (0.16)	-0.11 (0.15)	-0.44 (0.41)
<u>≥</u> 2 Weeks (%)	-1.44 (1.69)	-0.39 (0.63)	-0.16 (0.62)	-1.44 (1.69)
F-Tests of Pre-Tre	nds			
Any (0/1)	-2.77 (2.92)	-0.02 (0.03)	-0.02 (0.03)	-2.77 (2.92)
Count (#)	-0.36 (0.42)	0.00 (0.00)	0.00 (0.00)	-0.36 (0.42)
<u>></u> 2 Weeks (%)	-1.31 (1.75)	-0.01 (0.02)	-0.01 (0.02)	-1.31 (1.75)
Model Properties				
Covariates		Х		Х
Specification	TWFE	TWFE	Gardner	Gardner
Fixed Effects	ZIP code,	ZIP code,	ZIP code,	ZIP code,
	quarter-year	quarter-year	quarter-year	quarter-year
Clustering	County	County	County	County
* p < 0.05 ** p < 0.	01 *** p < 0.001	· · · · ·	¥	· · · ·

Appendix Table 5. Baseline Regression Results Excluding Manhattan

New York licensed medical cannabis dispensary data used in Drake et al. (2024) may be accessed at: https://github.com/cdrake219/NYdisp/tree/main