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Intended and Unintended Effects of Youth Bicycle Helmet Laws

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ABSTRACT

Over 20 states have adopted laws requiring youths to wear a helmet when riding a bicycle. We confirm previous research indicating that these laws reduced fatalities and increased helmet use, but we also show that the laws significantly reduced youth bicycling. We find this result in standard two-way fixed effects models of parental reports of youth bicycling, as well as in triple difference models of self-reported bicycling among high school youths that explicitly account for bicycling by youths just above the helmet law age threshold. Our results highlight important intended and unintended consequences of a well-intentioned public policy.

Keywords: bicycle helmet laws, bicycle fatalities, bicycle helmet use, bicycling, quasi-experiment

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1. Introduction

Every year, emergency departments in the United States treat hundreds of thousands of bicycle-related injuries, with hundreds resulting in deaths – usually due to head injuries (Rodgers 2000). Some large fraction of these deaths would likely have been preventable if the bicyclist had been wearing a properly fitted bicycle helmet, as there is ample medical evidence that helmets reduce the likelihood of serious head trauma and brain damage in bicycle accidents by as much as 85 percent, particularly among children (see, for example, Thompson, Rivara, and Thompson 1989). Despite this evidence, however, rates of bicycle helmet use are low, particularly among youths: about a third of youths age 5-15 in the United States who rode a bicycle in the past 12 months did not wear a helmet; helmet use among older adolescents is particularly low. When asked why they do not wear bicycle helmets, youths commonly complain about the “look” and “fit” of bicycle helmets. A common complaint about the former is that bicycle helmets make youths look like “geeks” or “nerds”, while “fit” problems generally include laments that helmets are too tight and induce sweating (AAA 1995).

Because the technological efficacy of bicycle helmets is well-accepted in the medical community, state and local governments have reason to try to increase helmet use. As with many public health policies, state efforts in this area have reflected two distinct types of approaches: “carrots” and “sticks”. Examples of the former include: bicycle helmet give-aways, educational outreach about the life-saving effects of helmets, and media campaigns designed to change social norms regarding bicycle helmets (e.g. to make it seem “cool” to wear a bike helmet). The most direct example of the “stick” approach has been to adopt laws that require youths to wear a helmet when riding a bicycle, a policy adopted by 21 US states over the past 15 years. Figure 1 shows the geographic distribution of helmet laws as of 2005. Although helmet

laws are largely coastal phenomena, over half of the US youth population is now covered by a state bicycle helmet law. Most of these laws apply to youths under age 16 and impose modest penalties for violations (e.g. verbal warnings, counseling, or a small fine). A recent quasi-experimental study returned evidence that helmet laws significantly reduced juvenile bicycle fatalities over the 1990s by about 11 percent (Grant and Rutner 2005).

How did helmet laws achieve these fatality reductions? One direct possibility documented in previous public health research is that helmet laws increased youth helmet use. Since there is general agreement in the medical community on the technological efficacy of helmets at reducing serious brain injuries, an increase in helmet use is one plausible and intuitively appealing mechanism (for reviews of the medical literature on the lifesaving effects of helmets, see Towner et al. 2002 and Karkhaneh et al. 2006). Despite this, the literature has not directly tested an alternative possibility: that mandatory helmet laws have reduced youth bicycling by increasing its costs.¹ Such a direct reduction in bicycling exposure would also be expected to reduce fatalities even in the absence of any changes in youth helmet use.

In this paper we provide the first comprehensive quasi-experimental estimates of the effects of mandatory bicycle helmet laws in the US on bicycling fatalities, helmet use, and bicycling behaviors among youths. Our main empirical approach is based on the differential timing of adoption of helmet laws across states, effectively identifying the helmet law effects from within-state changes in outcomes for residents of states adopting laws compared to the associated within-state changes in outcomes for youths in places that did not adopt a law in that same year. We first confirm and extend previous fixed-effects research on bicycling fatalities by showing that helmet laws significantly reduced bicycling fatalities among youths age 0-15 (i.e.,

¹ We are not the first to consider the unintended consequences of safety regulations. Peltzman (1973) points out that policies designed to improve automotive safety may lead to more careless or aggressive driving.

youths who were directly treated by most states' age-16 helmet laws) by about 19 percent. As expected, similarly specified fatality models among 16-30 year olds (who were not directly affected by the helmet laws) return much smaller and statistically insignificant effects of helmet laws. We also provide the first quasi-experimental evidence in the literature that an important mechanism behind the fatality reductions was an increase in helmet use: fixed-effects models using parental reports of child helmet use suggest that helmet laws significantly increased youth helmet use by 29-35 percent. Similarly specified models using self-reports of high school age youths also return evidence that helmet laws increase helmet use by 10-23 percent, though these estimates are less precise.

Our main contribution is to show that in addition to the increase in helmet use, there is also robust evidence for an unintended and previously undocumented mechanism: helmet laws produced modest but statistically significant reductions in youth bicycling participation of 4-5 percent. As with our helmet use estimates, we find this result in fixed-effects models of both parental reports of child bicycling and in self-reports of high school youths' bicycling, and the bicycling effects are statistically significant in both datasets. Moreover, our analysis of the reduction in bicycling goes further by showing that this main result also obtains in triple difference models that identify the effects of helmet laws by comparing changes in outcomes for youths under each state's helmet law age threshold (i.e., youths who were directly treated by the laws) with the associated changes in outcomes for youths at or above each state's helmet law age threshold (i.e., youths who were not directly treated by the laws) coincident with policy adoption. Estimates from these fully saturated models (with controls for age by state, age by year, and state by year fixed effects) provide particularly strong evidence that the significant reductions in bicycling associated with helmet law adoption are robust. Overall, our results highlight that a

full cost benefit analysis of the laws should take into account the previously ignored reductions in bicycling associated with helmet law adoption. More generally, our results highlight both the intended and the unintended consequences of a well-meaning public health policy.

The paper proceeds as follows: Section 2 outlines the data and empirical approach, Section 3 presents the results, and Section 4 concludes.²

2. Data Description and Empirical Framework

We use several different data sources to estimate the effects of mandatory helmet laws on youth outcomes. We first revisit the relationship between helmet laws and bicycling fatalities by updating Grant and Rutner's (2005) estimates with data from 1991-2005 from Fatality Analysis Reporting System (FARS).³ FARS contains data on a census of bicycle fatalities that involve an accident with a motor vehicle within the 50 States and the District of Columbia. Using these data, we create counts of the number of fatalities of bicycle riders among youths under age 16 (since most state helmet laws apply to children under age 16) and age 16-30 (as a specification check) in each state and year from 1991 through 2005.

² We do not provide a systematic review of the large public health and medical literatures that have examined the effects of bicycle helmet laws and bicycle helmets on helmet use, injuries, and bicycling using physical observations of cyclists and/or analyses of cross-sectional data. Multiple reviews of the literature have found that helmet laws have increased helmet use among the groups targeted by the laws (see, for example, Attwell et al. 2001, Towner et al. 2002, and Karkhaneh et al. 2006). Very few studies examine bicycle helmet laws in the United States; most research focuses on Australia or Canada (see, for example, Robinson 2003). An exception is Rodgers (2002) who used a telephone survey of approximately 900 young cyclists across several states that did and did not have helmet laws. That study found that, after controlling for individual demographic characteristics, state helmet laws were associated with statistically significant increases in the fraction of youths reporting that they "always" or "almost always" use a helmet when riding a bike, with effect sizes on the order of 18 percent. Notably, this study did not include non-cyclists, and as such could not address whether bicycling rates were different in states that had helmet laws compared to states that did not have these laws. Also, because the Rodgers (2002) study used a cross-sectional design, it could not address concerns about unobserved state specific characteristics that are correlated with both the presence of a helmet law and higher rates of helmet use.

³ From 2000-2005, four additional states adopted youth bicycle helmet laws.

To model the count nature of the outcome variables we estimate negative binomial models on annual state-specific fatality counts (Cameron and Trivedi 1998).⁴ In these models we control for average per capita income and the state unemployment rate, and to account for exposure we also enter the log of vehicle miles traveled per year in each state and the log of the relevant state/year population.⁵ Our policy variable of interest is equal to one in states and years when any helmet law for youths under age 16 is in place.⁶ To account for other state policies that may have affected crash risk and driving and bicycling opportunities, we also controlled for variables representing the presence of a state graduated driver licensing program with an intermediate phase, the presence of a primary enforcement seatbelt law, the presence of a secondary enforcement seatbelt law, the presence of a Zero Tolerance drunk driving law, the presence of a .08 BAC per se drunk driving law, and the presence of various speed limits (65mph and 70mph or greater) pertaining to cars on rural interstates. For all policy variables – including the helmet law variable – we control for the fraction of the year that a policy was in effect. The models also control for state and year effects, such that the coefficient of interest on the helmet law variable is identified from within-state changes in fatalities coincident with adoption of a helmet law, controlling for the associated changes in fatalities for residents of states that did not adopt a helmet law in the same year. Throughout, we cluster standard errors at the state level (Bertrand, Duflo, and Mullainathan 2004).

⁴ We used STATA 10.1's nbreg procedure to estimate a fixed effects negative binomial model for the fatality analysis and interpret the coefficients as marginal effects. Although fixed effects negative binomial models can yield inconsistent parameter estimates due to the “incidental parameters” problem (Hausman, Hall, and Griliches 1984), fixed effects poisson models (which do not suffer from this problem) returned nearly identical results to those presented here (Cameron and Trivedi 1998). Moreover, our fixed effects negative binomial model is supported by recent research (Allison and Waterman 2002).

⁵ Vehicle miles traveled come from the 1991-2005 issues of the US Department of Transportation publication *Highway Statistics*.

⁶ Alternative fatality specifications that control for the relevant fraction of 0-15 year olds who should have been directly affected by each helmet law (e.g., defining the helmet law variable to equal 11/15 for laws that apply only to youths under age 12) returned very similar results.

Our data on bicycling behaviors and helmet use for youths are from parental reports of child outcomes from the Centers for Disease Control's (CDC) Behavioral Risk Factor Surveillance System (BRFSS) over the period 1995-2000. These data include parental reports of bicycling behaviors and helmet use for youths as young as age 5. The BRFSS are random digit dialing telephone surveys, and adult respondents are asked a series of questions about the safety behaviors of the oldest child in their household between the ages of 5 and 15. Specifically, adults are asked: "How often within the last year did [Child's Name] wear a helmet when riding a bicycle?" Response options included: "Never wears a helmet", "Rarely wears a helmet", "Sometimes wears a helmet", "Almost always wears a helmet", and "Always wears a helmet".⁷ Importantly, adults can also indicate that the youth never rides a bicycle. We exclude a small number of adults who report that they don't know or refused a response to the bicycle helmet question, and we restrict attention to adults with no missing demographic information. We use responses to this question to create a variable called Bicycle Rider that equals one if the parent chose any of the latter five responses. To measure helmet use, we follow previous research by creating a measure called Frequent Helmet Use that equals one if the parent reports that the youth wore a helmet either "almost always" or "always". We also created a continuous helmet use measure that equals 1 for "always" wears a helmet, .75 for "almost always", and so forth through 0 for "never" wears a helmet.

We supplement the parental reports of child bicycling behaviors using restricted-use versions of the national Youth Risk Behavior Surveillance System (YRBSS) over the period 1991-2005. These surveys are coordinated every other year by the Centers for Disease Control and are administered by paper and pencil to high school students at schools in the spring. These

⁷ In the BRFSS data, the question about helmet use was in an Injury Control topical module that was only asked in a small handful of states for the even-numbered years between 1995 and 2000. For the odd-numbered years, however,

data provide standard demographic characteristics, information on bicycle riding and helmet use, and state of residence (provided by the CDC in a confidential request). For the YRBSS analysis we restrict attention to youths with no missing data on demographic characteristics or helmet use.⁸

Like the BRFSS, the core YRBSS questionnaire in each year since 1991 has included the following question: “When you rode a bicycle during the past 12 months, how often did you wear a helmet?”⁹ Response options include: “Did not ride a bicycle”, “Never”, “Rarely”, “Sometimes”, “Most of the time”, and “Always”.¹⁰ We consider the same outcomes as in the parental reports for the youth self-reports: Bicycle Rider and Frequent Helmet Use (“always” or “most of the time” wore a helmet).

To estimate the effect of the mandatory helmet laws in the BRFSS and YRBSS data, we first follow the basic approach in the fatality model described above and estimate two-way fixed effects models on the group targeted by most helmet laws – those under age 16 in the YRBSS (or, alternatively, parents of youths under age 16 in the BRFSS).¹¹ The two-way fixed effects model for 15 year olds and younger in both datasets amounts to OLS estimation of the following:

$$(1) Y_{ist} = \beta_0 + \beta_1 X_{ist} + \beta_2 (\text{Mandatory Bicycle Helmet Law})_{st} + \beta_3 Z_{st} + \beta_4 S + \beta_5 T + \epsilon_{ist}$$

we observe large samples of adults across all 50 states and the District of Columbia.

⁸ Notably, the national YRBSS survey was not explicitly designed to produce estimates that are representative at the state level. Despite this, previous researchers have used these data in state policy evaluations such as ours (see, for example, Gruber and Zinman 2001, Carpenter and Cook 2008, Carpenter and Stehr 2008, and others).

⁹ In 1991-1997, the core YRBSS questionnaire also included a lead-in question about intensity of bicycle riding in the previous year. Specifically, youths were asked: “During the past 12 months, how many times did you ride a bicycle?,” with response options including: 0 times, 1 to 10 times, 11 to 20 times, 21 to 39 times, and 40 or more times. Unfortunately, over this shorter sample period we do not have enough state policy changes or sufficient statistical power to estimate precise helmet law effects. Models that were specified in a similar fashion to our main estimates returned results that were qualitatively similar to those we present below (i.e. state helmet laws were associated with reductions in youth bicycling intensity).

¹⁰ Note that these response options are worded in a slightly different fashion than those on the BRFSS questionnaire, though we treat them in a similar fashion.

¹¹ Note that the BRFSS does not ask parents about outcomes for children age 16 and older.

where Y_{ist} are the outcomes of interest (Frequent Helmet Use, Continuous Helmet Use, and Bicycle Rider).¹² X_{ist} is a vector of individual student characteristics that includes: age dummies, grade dummies, and race dummies in the YRBSS. In the BRFSS, X_{ist} is a vector of parental characteristics that includes: age dummies, marital status, and race dummies. Z_{st} is a vector of other potentially relevant state policies and characteristics, including: the state unemployment rate, a dummy for the presence of a mandatory seatbelt law, and a dummy for any state graduated driver licensing program with an intermediate phase. Mandatory Bicycle Helmet Law is an indicator variable equal to one if the respondent lives in a state that had a mandatory helmet law over the previous 12 months.¹³ The coefficient of interest, β_2 , captures the relative effect of the mandatory helmet law on youth outcomes by comparing within state changes in outcomes for youths in helmet adopting states coinciding with the law taking effect to the associated changes in outcomes for youths in states that did not experience a policy change in that year. We include unrestricted state and year fixed effects, and we cluster standard errors at the state level.

To complement the difference in differences specifications in the YRBSS data we also explicitly take advantage of the age threshold in the helmet law statutes to estimate difference in difference in differences (DDD) models (Gruber 1994). Specifically, we use changes in bicycling for youths at or above the age cutoff coincident with helmet law adoption as controls for the associated changes in outcomes for youths below the age cutoff. In practice, this amounts

¹² Probit models returned very similar results.

¹³ This time window is chosen to account for the wording of the question. Note that in theory we would want to control for the fraction of the previous year the youth was covered by a helmet law. This is made difficult by the fact that we do not observe the actual date the surveys were administered; we simply know they were done in the spring. We choose to drop from the sample any observations from a state/year combination where this ‘fractional’ treatment status is uncertain. In practice, the timing of the effective dates of the laws combined with the fact that the surveys are only done in odd-numbered years means that this excludes very few observations. Creating such a ‘fractional’ helmet law variable with the assumption that all youths are surveyed in, say, March, returned very similar results. In the BRFSS analyses we are able to account for a handful of state laws that set the relevant helmet law threshold at age 12 instead of age 16 due to the inclusion of parental reports for children as young as age 5. In

to estimating a variant of equation (1) on the full sample of youths, augmented by including interactions for a “treatment group” dummy (equal to one for all youths under the state’s helmet law threshold) with each state dummy and each year dummy. Importantly, this model also supports a full set of dummies for each state/year interaction. The key advantage to this model is that the only assumption required for identification of the helmet law effect is that there are no shocks to outcomes in experimental states that affect youths under the age threshold differently than youths over the age threshold. We implement this model by estimating the following equation:

$$(2) Y_{ist} = \beta_0 + \beta_1 X_{ist} + \beta_2 (\text{Mandatory Bicycle Helmet Law} * \text{Treatment Group})_{st} + \beta_3 S + \beta_4 T + \beta_5 (\text{Treatment Group} * T) + \beta_6 (\text{Treatment Group} * S) + \beta_7 (S * T) + \varepsilon_{ist}$$

where all variables are as defined above. Treatment Group is an indicator equal to one if the youth is below her state’s relevant helmet law threshold. Note that in this model Z_{st} , the vector of state characteristics (such as the unemployment rate) that vary only at the state/year level, falls out of the above specification since a full set of state by year interactions are included.

3. Results

We begin by confirming and extending the result in Grant and Rutner (2004) that mandatory bicycle helmet laws reduced bicycling fatalities. Recall that Grant and Rutner found that helmet laws reduced juvenile fatalities by about 11 percent over the period 1990-2000. We present the main results on bicycle fatalities among youths age 0-15 in the FARS data from 1991-2005 in Table 1. Each entry in the table shows the coefficient and standard error on the Mandatory Bicycle Helmet Law indicator from estimation of the fixed effects negative binomial model of

all the BRFSS models the Helmet Law variable is coded as “one” when the referenced youth is covered and “zero” otherwise.

bicycle fatalities. These coefficients can be interpreted as percent changes in bicycling fatalities associated with adoption of a helmet law. Column 1 presents the results for youths who should have been most affected by helmet laws (youths age 0-15) while Column 2 presents the results for young adults who should have been unaffected by the policies (age 16-30). The findings in Table 1 confirm the key results from Grant and Rutner (2004) for the 1990s: state bicycle helmet laws were associated with statistically significant reductions in youth bicycle fatalities among youths age 0-15 on the order of 19 percent but were not significantly associated with bicycle fatalities among 16-30 year olds.¹⁴

What factors contributed to the bicycling fatality reductions associated with helmet law adoption documented in Table 1? We investigate this issue using helmet use and bicycling behaviors from the BRFSS and YRBSS. We begin by presenting key variable means for the parental reports (BRFSS) in Table 2 and the youth self report data (YRBSS) in Table 3. The BRFSS data reveal that about 84 percent of parents report their oldest child age 5-15 rode a bike in the previous year, and about two thirds of bicycle riders wore a helmet when riding. The YRBSS means in Table 3 show that 71 percent of high school age youths report that they rode a bike in the previous year. Among past year high school-age cyclists in the YRBSS, however, helmet use is rare: only about 16 percent of these youths report that they wore a helmet when riding a bike. The differences in helmet use across the two datasets are largely a function of the

¹⁴ Our estimate of the effect of helmet laws on bicycling fatalities among youths is larger than that reported in Grant and Rutner (19 versus 11 percent). Personal correspondence with Darren Grant indicates that they examine deaths of all juveniles age 0-17 as their outcome variable of interest. We choose to focus on 0-15 year olds since the vast majority of state bicycle helmet laws explicitly apply only to those under age 16. Since we consider 16 and 17 year olds to be untreated by the laws, our estimate is less attenuated (i.e. larger). Indeed, when we estimate the same specification on deaths among 0-17 year olds, our estimate is 13 percent, considerably closer to that reported in Grant and Rutner (2004). We revisit the magnitude of the effect below, but for comparison purposes note that Dee (2009) finds that motorcycle helmet laws reduced motorcycle fatalities by about 27 percent over roughly this same time period.

younger age distribution of children in the BRFSS, since bicycling and helmet wearing rates strongly decline with age in this range.

We provide direct evidence on the effects of helmet laws on helmet use in Table 4. Specifically, we present the coefficients on the helmet law variable from two-way fixed effects models estimated on the full sample (i.e., not conditional on past year bicycle riding) that control for individual demographic characteristics, time-varying state characteristics, and a full set of state and year fixed effects. We present the results for the parental reports of child helmet use behaviors from the BRFSS in Column 1, and we present the associated results using the high school youths' self reports from the YRBSS in Column 2. In the top panel we present estimates from a linear probability model where the outcome is an indicator for whether the youth always or almost always wore a helmet when riding a bike. We complement these results in the middle panel by presenting models where the outcome is a continuous measure of helmet use (where "never" wore a helmet = 0 and "always" wore a helmet = 1). Finally, in the bottom panel we show the two-way fixed effects results for the bicycling participation outcome that equals 1 if the youth rode a bike in the past year. Each entry in the table reflects the coefficient on the helmet law indicator, and reported standard errors have been clustered by state.

The results in Table 4 show that helmet laws did, in fact, increase helmet use. This is particularly evident from the parental reports of child bicycling behaviors in Column 1 as reported in the BRFSS. In the top panel of Column 1, for example, we estimate that adoption of a helmet law increased frequent helmet use by 10.7 percentage points, or about 35 percent relative to the pre-reform mean. We find a much smaller point estimate in Column 2 for the high school youths' self reports, though notably these are off a much smaller base, such that the estimated proportional effect is nontrivial (about 10 percent). Results for the continuous measure

of helmet use in the middle panel of Table 4 return more consistent evidence across the two columns that helmet laws increased helmet use by 29 and 23 percent relative to pre-reform levels in the BRFSS and YRBSS, respectively.

In the bottom panel of Table 4 we also present the first evidence for our main contribution: namely, that helmet laws also reduced youth bicycling. Specifically, we find in both the parental reports in Column 1 and in the high school youth self reports in Column 2 that helmet laws reduced the probability a youth rode a bicycle in the previous year by 3-4 percentage points, and both estimates are statistically significant at the five percent level. Relative to the pre-reform means, the estimates in Columns 1 and 2 of the bottom panel of Table 4 suggest helmet law effects on bicycling of about 4 and 5 percent, respectively.¹⁵

We further investigate the robustness of the main bicycling result for high school youths in Table 5. The top panel of estimates shows results for the two-way fixed effects models, while the bottom panel shows results for the augmented triple difference models with a full set of age by state, age by year, and state by year fixed effects. The models in the bottom panel identify the helmet law effects in youth bicycling by comparing the associated outcomes for high school youths under the helmet law age threshold with those for youths at or above the age threshold (who were therefore untreated) coincident with policy adoption, and we show the estimate on the

¹⁵ In results not reported but available upon request we found a small and statistically insignificant association between state graduated driver licensing (GDL) programs and youth bicycling in the YRBSS. This null finding is important, since it is inconsistent with a plausible confounding story: namely, that changes in driving availability (which are directly regulated by GDL program adoptions) might induce a substitution across modes of transportation that would bias our main bicycling result. See, for example, Dee et al. (2006) for a description of graduated driver licensing programs. In addition to the null empirical finding on GDL in the bicycling regression, it is also helpful to note that GDL programs were much more widespread than helmet laws over this time period. Moreover, GDL programs do not strictly adhere to the over/under 16 distinction in the same way that helmet laws do. This is because although GDL programs have common features (such as requiring intermediate phases for obtaining a driver's license), states are still free to set the minimum age at which one can obtain her learner's permit. This varies substantially across states, which further casts doubt on the ability of driving availability to explain our main bicycling result. Because we do not observe the respondent's exact age, controlling for when the state adopted a GDL program is the best we can do to control for driving availability. Unfortunately, the YRBSS data do not contain consistent questions about driving behaviors or license status.

coefficient of interest from equation (2) on the helmet law * treatment group interaction. In the top panel of Column 1 of Table 5 we reprint the baseline DD estimate from the bottom panel of Column 2 of Table 4 that helmet laws reduced bicycling among youths age 15 and under by a statistically significant 3.8 percentage points. In the bottom panel of Column 1 of Table 5 we show the associated DDD estimate for the full YRBSS sample which confirms that the DD finding on bicycling we saw in Table 4 is robust to incorporating the bicycling behaviors of slightly older youths. Specifically, our DDD estimate suggests that helmet laws reduced bicycling by a statistically significant 3.1 percentage points.

The remaining columns of Table 5 confirm that the YRBSS bicycling reduction is robust to other reasonable changes in specification and sample. In Column 2 we restrict attention to 15 and 16 year olds, who should be the most directly comparable, and we continue to find that helmet laws significantly reduced youth bicycling. In the top panel of Column 3 we estimate a DD model that in addition to the usual helmet law indicator also includes a variable that “turns on” two years prior to each state’s actual helmet law. The use of this lead variable captures the possibility that helmet laws were adopted in response to some systematic change in bicycling behaviors within adopting states, potentially affecting the interpretation of our helmet law effects. Despite this possibility, we find that including the two year lead variable does not change our main result that helmet laws reduced bicycling, and the coefficient on the lead variable is small and statistically insignificant. Column 4 restricts attention to youths living in states that ever adopted a helmet law, another standard robustness check. This approach returns similar DD and DDD estimates to the baseline, both of which are statistically significant. This confirms that our bicycling estimates are not driven by offsetting changes in behavior in control states coincident with helmet law adoptions.

Columns 5-7 of Table 5 address concerns about potential outliers and the geographic distribution of helmet law states. Figure 1 shows the set of states that adopted helmet laws over our sample period. Notably, the states are largely concentrated on the coasts, with no states in the Midwest or Mountain areas that ever adopted statewide helmet laws. In Column 5 we estimate models on a sample that drops all states in divisions that contain no helmet law adopters (East North Central, West North Central, and Mountain). The intuition here is that the set of “control” states is arguably more like the set of “treatment” states when we restrict attention to states within the same Census divisions; doing so has little effect on the bicycling estimates, and they remain statistically significant. In Columns 6 and 7 we drop California and Florida, respectively, from the sample. Each of these states adopted a helmet law over this time period and contributes over 1,500 observations to the full sample. Ensuring that these large states are not driving the bicycling estimates – as confirmed in Columns 6 and 7 – is an important robustness check. In results not reported, we also found that excluding each other helmet law adopter individually did not materially alter our main finding that helmet laws reduced youth bicycle riding in the YRBSS.¹⁶

¹⁶ We also performed these same robustness checks on the fatality models presented in Table 1 and obtained similar results. In results not reported but available upon request we also examined a number of other “placebo” outcomes that should not have been directly affected by helmet laws (e.g., seatbelt use, motorcycle riding, smoking, drinking, sexual activity) to ensure that our observed effects are not simply proxying for contemporaneous changes in (un)healthy behaviors that – while unlikely in our empirical framework – could bias our estimated helmet law effects. We found no consistent evidence that bicycle helmet law adoption was associated with economically or statistically significant effects on any of the other risky behaviors. These null findings further reinforce our main result that helmet laws reduced bicycling. Finally, we also examined the possibility that helmet laws – by reducing bicycling – could have resulted in other negative health spillovers such as lack of exercise and increases in obesity. Unfortunately, we do not observe consistent measures of body weight or sedentary activity (such as hours of television viewing) over the entire sample period 1991-2005. Using the same empirical approach as in Table 5, however, we found no evidence that helmet laws reduced physical activity as measured by: 1) participation in hard physical exercise for at least 20 minutes in any day in the previous week; 2) participation in muscle strengthening activities (e.g. weight lifting or push-ups) on any days in the previous week; 3) exercising for the purpose of losing weight or maintaining current weight; and 4) participation in any sports teams. These results are available upon

4. Discussion and Conclusion

The results above show that state laws adopted over past two decades that require youths to wear helmets when riding a bicycle reduced youth bicycling fatalities by about 19 percent, increased helmet use by 20-34 percent, and (unintentionally) reduced bicycling by 4-5 percent. Given that the foregone utility of riding a bicycle would seem relatively substantial, why do helmet laws lead to reduced cycling? There are several possibilities. First, the costs of helmet use are likely nontrivial for some youths. In addition to the direct monetary costs of bicycle helmets (usually between \$10 and \$40), there are likely to be substantial social costs (recall that survey evidence shows that youths do not like wearing helmets primarily because they are “uncool”). These costs are likely to be magnified if there are significant peer effects. Second, there is evidence that youths have suboptimally high discount rates (Gruber 2001), such that some youths might place too little weight on the expected gain in future utility from the prevention of injury or death relative to the costs of wearing helmets today. Finally, note that there are several reasonably close alternatives to bicycling (such as skateboarding and in-line skating) that are not regulated in the same way with respect to mandated helmet use. Helmet laws change the relative prices among these activities, such that the robust reduction in bicycling we observe may be at least partially offset by increases in other related activities.¹⁷

Are the magnitudes of our results plausible given what we know about the technological efficacy of bicycle helmets and the helmet law-induced changes in helmet use? To understand what fatality reduction would be implied by our estimated increases in helmet use, we use Evans’ basic calculation for seat belt effectiveness that uses information on the technological efficacy of

request and suggest that although helmet laws reduced bicycling, they did not appear to have large spillovers to other measures of physical activity for youths.

¹⁷ Unfortunately, we do not observe measures of participation in these activities in our data, nor can we credibly identify fatalities from these activities in the FARS in order to directly test this risk compensation hypothesis.

helmets (1987). This calculation shows that: $F = [E \Delta h] / [1 - Eh]$, where F is the proportional fatality reduction, E is the technological efficacy of bicycle helmets, and h is the helmet use rate. Note that about 66 percent of our 5-15 year old sample from the BRFSS wears helmets, and Table 4 showed that helmet laws increased helmet use by about .10 (using the continuous helmet use measure). We are not aware of estimates from the public health or medical literatures that directly estimate the technological efficacy of bicycle helmets at reducing fatalities, though Thompson et. al. (1989) use case-control methods to find that bicycle helmets reduce serious head and brain injuries by over 80 percent; they suggest the efficacy is likely even greater for youths. Perhaps the closest comparison regarding helmet efficacy in the context of fatalities comes from Dee's motorcycle helmet study which uses Evans' "double pairs" comparisons to find that motorcycle helmets reduce motorcycle fatalities by 34 percent (2009). If bicycle helmets were equally effective as motorcycle helmets, we would expect helmet laws to reduce bicycling fatalities by about 4.4 percent $[(.34*.1)/(1-.34*.66)]$ in the absence of any reduction in risk exposure.

Our fatality estimate of 19 percent is substantially larger than this, though several points are worth mentioning. First, our fatality estimate has a relatively large standard error, such that we cannot rule out that the true fatality effect is substantially smaller than our reported estimate. Second, the 4.4 percent implied fatality reduction does not take into account the direct reduction in risk exposure associated with reduced bicycling which should independently reduce fatalities.¹⁸ Third, if bicycle helmet laws increase police attention to bicycling routes, then we

¹⁸ Although we estimate that bicycling participation fell by about 5 percent, it is likely that overall bicycling miles travelled fell even more. Unfortunately, neither the YRBSS nor the BRFSS asked consistent questions about bicycling intensity over the full sample period. As noted previously, however, the YRBSS did ask about the number of instances of bike riding from 1991 to 1997. We estimated equation (1) on this outcome (using the midpoints of the ranges and coding the top category as 50 instances) and found that helmet laws reduced bicycling among high school youths age 15 and under by 2.34 instances, or about 11 percent relative to the pre-reform mean of 21.32 instances. This suggests that the true overall reduction in bicycling miles travelled – and thus exposure to potential

would expect an additional fatality reduction associated with more careful driving along those enforced routes that has nothing to do with helmet technology or helmet use per se. Fourth, parents may purchase other safety equipment such as blinking lights or reflective clothing for their children at the same time that they might buy their child a helmet to comply with a newly passed helmet law. If such equipment actually prevents accidents or reduces their severity, then its purchase could substantially augment the effectiveness of helmet laws. Finally, our estimated fatality effect size is comparable in magnitude to credible estimates from public policy in related settings; multiple recent evaluations of motorcycle helmet laws find that they reduce motorcycling fatalities by 27-33 percent (see, for example, Dee 2009, Houston and Richardson 2008). Moreover, as in our setting for bicycle helmet laws, Dee shows that motorcycle helmet laws induce larger changes in fatalities than can be directly implied by reasonable changes in helmet use and estimates of helmet efficacy.

The current study has several important limitations. First, we did not control for the arguably important effects of enforcement, outreach, or media campaigns because we are not aware of a source that consistently tracks these efforts at the state level over our sample period. Efforts of this type such as “helmet giveaways” are of concern if they are correlated with adoption of a mandatory helmet law. Notably, however, this type of bias is not likely to be the main determinant of our estimated effects on bicycling behavior, since helmet giveaways should, if anything, increase bicycling behavior. Given that we find significant *decreases* in bicycling participation, this suggests that our estimates on bicycling are likely understated with respect to outreach campaigns. Second, all of the bicycling data are self-reported, either by parents (in the BRFSS) or students (in the YRBSS). While we find it plausible that youths and parents might be

bicycling accidents – is larger than our bicycling participation estimates in Tables 3 and 4. These estimates are of course based on fewer state changes and as such are less precise than the results for bicycle riding, which we

more likely to falsely report helmet use in the presence of a helmet law (perhaps because stigma associated with not wearing a helmet has increased), this type of bias is unlikely to affect our estimates on bicycling participation. Third, we lack important data that would allow us to provide a more comprehensive assessment of helmet laws. Information on helmet ownership, for example, would be extremely useful since research has shown a strong concordance between ownership and use (Schieber et. al. 1996). This information could be important for understanding the likely relative effectiveness of helmet “give-aways” versus media campaigns to reduce the stigma of wearing a helmet (i.e., if the main effect is driven by helmet availability, give-aways could be particularly effective). Finally, information on other adverse outcomes (such as head injuries, emergency room visits, and hospitalizations) would be necessary for a complete accounting of the costs and benefits of helmet laws.

Despite these limitations, our research has provided the first comprehensive quasi-experimental evaluation of the effects of helmet laws on bicycling and helmet use behaviors among youths. As other states consider helmet laws as a way to reduce bicycling related injuries and fatalities, they should keep in mind that although the laws increase helmet use and reduce fatalities, they are also likely to reduce bicycling among the targeted group. While it is possible the public health benefits from mandating helmet laws may outweigh the reductions in utility associated with less bicycling, future research evaluating the full costs and benefits of these policies should acknowledge these effects.¹⁹

observe over the entire sample period from 1991 to 2005 in the YRBSS.

¹⁹ We provide here a very rough back of the envelope calculation of these effects. In 2005, there were 103 bicycle fatalities among 0-15 year olds in the 36 states without a helmet law applying to youths under age 16. We estimate that helmet laws would reduce these fatalities by 19 percent, or 19.57 bicycling fatalities per year. Using a value of a statistical life of about \$8 million in 2005 dollars (Viscusi and Aldy 2003), the annual benefit from saving these additional lives would be roughly \$157 million (\$8 million * 19.57 lives). There are about 41.9 million youths age 0-15 in these states without helmet laws. Given that 84 percent of youths age 5-15 ride bicycles and 35 percent of these youths do not wear helmets, we estimate that a helmet law would potentially constrain about 12.3 million youths (41.9 million * .84 *.35). Alternatively, if we consider only those youths whose helmet use increases as

having been actually constrained, then we estimate that helmet laws constrain about 3.5 million youths (41.9 million * .84 * .10). Assuming a real discount rate of 4 percent and a 30-year time horizon, we estimate the present discounted value of the benefit to be roughly \$800 for each cyclist whose helmet use increases in response to the law. Note this calculation ignores other benefits from the laws (such as injury reductions) as well as other costs (such as the value of foregone cycling).

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Table 1:
Bicycle Helmet Laws Reduced Bicycling Fatalities
State and Year Fixed Effects Models, FARS 1991-2005

| | (1) | (2) |
|------------|--------------------|-----------------|
| | 0-15 year olds | 16-30 year olds |
| Helmet Law | -.189*** (.069) | -.057 (.078) |
| N | 765 | 765 |

Each entry represents a separate regression estimated on the state/year count of fatalities of cyclists in different age groups. We present marginal effects and associated standard errors clustered at the state level. Models also include controls for: the log of the relevant age-specific population in the state, the log of vehicle miles traveled in the state, average income in the state, the state unemployment rate, an indicator for the presence of a .08 BAC law, a Zero Tolerance law, a graduated driver licensing law, a primary enforcement seatbelt law, a secondary enforcement seatbelt law, and speed limits (65mph, 70+ mph). * significant at 10%; ** significant at 5%; *** significant at 1%.

Table 2:
Descriptive Statistics, Parental Reports, 1995-2000 BRFSS Data

| Variable | Mean |
|--|------|
| Oldest child age 5-15 in household rode a bike last year | .84 |
| How often child wore helmet, among riders: | |
| Always | .38 |
| Almost Always | .11 |
| Sometimes | .11 |
| Rarely | .06 |
| Never | .35 |
| Demographic Characteristics of the Responding Parent: | |
| Female | .54 |
| White, non-Hispanic | .79 |
| Age | 37.5 |

Weighted means.

Table 3:
Descriptive Statistics, High School Youth Self-Reports, 1991-2005 YRBSS Data

| Variable | Mean |
|---|------|
| Rode a bike last year | .71 |
| How often wore a helmet, among riders: | |
| Always | .04 |
| Almost Always | .04 |
| Sometimes | .04 |
| Rarely | .06 |
| Never | .83 |
| Lives in a state that adopted a helmet law over sample period | .41 |
| Individual is covered by a mandatory state helmet law | .13 |
| Female | .49 |
| Black | .14 |
| Other race | .10 |
| Hispanic | .11 |
| Grade 9 | .27 |
| Grade 10 | .25 |
| Grade 11 | .24 |
| Grade 12 | .24 |
| Age 14 | .10 |
| Age 15 | .24 |
| Age 16 | .26 |
| Age 17 | .25 |
| Age 18+ | .15 |
| Weighted means. | |

Table 4:
Bicycle Helmet Laws Increased Helmet Use But Decreased Bicycling
State and Year Fixed Effects Models, BRFSS (Column 1), YRBSS (Column 2)

| | (1) | (2) |
|--|---|--|
| | BRFSS parental reports, oldest child is 5-15 | YRBSS high school self-reports, youths age 15 and under |
| Always or almost always wears helmet | | |
| Helmet Law | .107*** (.027) | .006 (.020) |
| Implied effect size as fraction of pre-reform mean | + 34.9% | + 9.7% |
| R-squared | .228 | .055 |
| N | 115886 | 34014 |
| Continuous Bicycle Helmet Use (0=never, 1=always) | | |
| Helmet Law | .100*** (.024) | .019 |
| Implied effect size as fraction of pre-reform mean | + 28.7% | + 22.9% |
| R-squared | .259 | .080 |
| N | 115886 | 34014 |
| Rode a Bike | | |
| Helmet Law | -.030** (.015) | -.038*** (.013) |
| Implied effect size as fraction of pre-reform mean | -3.6% | - 4.7% |
| R-squared | .087 | .074 |
| N | 115886 | 34014 |

Sample in column 1 is BRFSS parental reports from 1995-2000. Sample in column 2 is YRBSS self-reports from 1991-2005. Each entry represents a separate regression estimated using OLS. Robust standard errors clustered at the state level are shown in parentheses. All models include state and year fixed effects and time-varying state characteristics, including: the state unemployment rate, a dummy for the presence of a mandatory seatbelt law, and a dummy for any state graduated driver licensing (GDL) program with an intermediate phase. Other individual demographic characteristics in the BRFSS models in the top row but not reported here include: parental age, parental age squared, six parental education dummies, parental gender, a white non-Hispanic dummy, dummies for each age of oldest child under 15, and survey month dummies. Other individual demographic characteristics in the YRBSS models in the bottom row but not reported here include: age dummies, grade dummies, and race dummies.
 * significant at 10%; ** significant at 5%; *** significant at 1%.

**Table 5:
The Effect of Bicycle Helmet Laws at Reducing Youth Bicycling is Robust
YRBSS 1991-2005**

| | (1) | (2) | (3) | (4) | (5) | (6) | (7) |
|---|--|---------------------------|---|-------------------------------------|--|-------------------|-------------------|
| | Baseline - corresponding to Table 5, Column 2 (top panel) and Column 4 (bottom panel)) | Only 15 & 16 year olds | Baseline, with a control for 2 year lead of helmet law | Baseline, adopter states only | Drop all states in East North Central, West North Central, and Mountain Divisions | Drop California | Drop Florida |
| Difference in Differences, age 15 and under | | | | | | | |
| Helmet Law – 2 year lead | | | .016 (.019) | | | | |
| Helmet Law | -.038*** (.013) | -.040*** (.014) | -.046*** (.010) | -.039** (.015) | -.046*** (.015) | -.041** (.015) | -.033** (.013) |
| R squared | .074 | .075 | .074 | .080 | .075 | .074 | .074 |
| N | 34014 | 24194 | 34014 | 14678 | 25794 | 29001 | 31979 |
| Difference in Difference in Differences, all youths | | | | | | | |
| Helmet Law * Under Helmet Law Threshold | -.031** (.015) | -.047** (.018) | -- | -.055** (.019) | -.033** (.016) | -.030 (.019) | -.037** (.017) |
| R squared | .099 | .092 | | .101 | .104 | .102 | .100 |
| N | 109804 | 52394 | | 45707 | 84388 | 94818 | 103645 |

See notes to Table 5.

**Appendix A:
State Bicycle Helmet Laws for Youths**

| State | Ages Covered | Effective Date |
|----------------------|--------------|----------------|
| Alabama | under 16 | 9/19/1995 |
| California | under 5 | 1987 |
| California | under 18 | 10/8/1993 |
| Connecticut | under 15 | 10/1/1993 |
| Connecticut | under 16 | 5/14/1997 |
| Delaware | under 16 | 4/1/1996 |
| District of Columbia | under 16 | 5/23/2000 |
| Florida | under 16 | 1/1/1997 |
| Georgia | under 16 | 7/1/1993 |
| Hawaii | under 16 | 1/1/2001 |
| Louisiana | under 12 | 3/1/2002 |
| Maine | under 16 | 9/18/1999 |
| Maryland | under 16 | 10/1/1995 |
| Massachusetts | under 5 | 1990 |
| Massachusetts | under 13 | 3/8/1994 |
| Massachusetts | under 17 | 11/25/2004 |
| New Hampshire | under 16 | 1/1/2006 |
| New Jersey | under 14 | 7/1/1992 |
| New Jersey | under 17 | 3/1/2006 |
| New York | under 5 | 1989 |
| New York | under 14 | 6/1/1994 |
| North Carolina | under 16 | 10/1/2001 |
| Oregon | under 16 | 7/1/1994 |
| Pennsylvania | under 5 | 1991 |
| Pennsylvania | under 12 | 2/1/1994 |
| Rhode Island | under 9 | 7/1/1996 |
| Rhode Island | under 16 | 1998 |
| Tennessee | under 12 | 1/1/1994 |
| Tennessee | under 16 | 2000 |
| West Virginia | under 15 | 7/1/1996 |

Information on helmet laws was determined through comparisons of multiple sources, including: the Bicycle Helmet Safety Institute, the Snell Memorial Foundation, Safe Kids USA, the Insurance Institute for Highway Safety, and various Lexis-Nexis and legislative document searches. Note that we do not use all of these helmet laws in all of the empirical work. For the YRBSS analyses, we use laws applying to high school age youths (14-18). For the BRFSS analysis, we use laws applying to youths age 5-15. In cases where we could not identify the exact effective date and where such a choice would affect our definition of treatment and control groups, we excluded the small fraction of youths for whom treatment status was uncertain.