

# **Bicycle Infrastructure and Commercial District Change**

03/30/2018

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Funded by Caltrans Grant #65A0529, UCCONNECT Task Order 052.2

## Acknowledgments

This research was funded by Caltrans and completed by the Center for Community Innovation.

Special thanks to Simon Hochberg for coding assistance and Brendan Hurley for graphics. Thanks also to research assistants Rebecca Farhi, Javier San Millan Tejedor, Copeland Laris, Sho Takiguchi, and Violet Sinnarkar.

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## Executive summary

Cities, counties, and states can provide bicycle infrastructure to prompt existing cyclists to bicycle more frequently and encourage would-be cyclists to make trips by bike. One relatively understudied aspect of bicycle infrastructure is its impact on business establishments. Advancing the understanding of this relationship is increasingly relevant as cities across California continue to push for more bicycle infrastructure projects—and face opposition from business constituencies, among others.

This research explores this relationship in two ways. First, we use secondary data on business performance to analyze the impact bicycle infrastructure has had on sales and business closures in two counties in the San Francisco Bay Area: San Francisco and Alameda Counties. Second, an intercept survey of shoppers on matched pair corridors helps to further illuminate the relationship between bicycle infrastructure, mode choice for shopping trips, and consumer behavior.

Does bicycle infrastructure impact business sales? It depends. In San Francisco and Alameda Counties, location on Class II facilities (dedicated bike lanes) was not associated with a significant change in sales, except for on neighborhood roads, which saw an increase. Further, automobile-oriented businesses and businesses selling home goods that were located on corridors with Class II facilities did see a decline in sales (of around \$100,000 per year each). Location on Class III facilities (shared roadways) in San Francisco had no significant impacts on sales for businesses. In Alameda County, location on Class III infrastructure was associated with a generally positive change in sales, though in the case of facilities on secondary roads, there was a negative association.

This generally disproves business owners' claims that bike infrastructure is bad for business, though it generally does not confirm cyclist advocates' claims that bike infrastructure is good for business. Instead, it appears that for businesses in San Francisco, there are a multitude of other factors that do have a determining effect on the change in sales over time. Business characteristics were overall the most reliable predictors of sales. Neighborhood characteristics were likewise poor predictors. Corridor characteristics were somewhat predictive, with primary roads associated with sales declines.

In terms of business turnover, both classes of bike infrastructure in San Francisco showed no significant relationship with the likelihood of turnover. Change in on-street parking was also not significant in the Class II model for San Francisco. In Alameda County, businesses abutting Class II and Class III corridors saw a lower likelihood of turnover. These models suggest that bicycle infrastructure either has no effect or a positive effect on business outcomes.

In sum, the models suggest that bicycle infrastructure has a generally mixed effect on the change in sales over time at an individual business level, with the most positive effects occurring on neighborhood roads. The relationship between business turnover and bicycle infrastructure is either neutral in the case of San Francisco or pro-business in the case of Alameda County, with location on bicycle infrastructure of all types resulting in a reduced likelihood of turnover.

Through a survey of shoppers on matched pair corridors in San Francisco and Alameda Counties, we failed to find a coherent distinction in shopper behavior on corridors with and without bicycle infrastructure. Similar rates of shoppers arriving by bike were registered for corridors with and without bicycle infrastructure. When asked about the general frequency with which they visit the corridor by bike for shopping trips, all respondents - those who arrived by bike and otherwise - reported coming to corridors without bicycle infrastructure more frequently to shop. It appears that cyclists in the survey sample were wealthier on the whole than shoppers who arrived by other modes and that any differences in shopping patterns were likely due to that difference, rather than the mode used or the presence of bicycle infrastructure.

Taken together, these findings indicate that overall, bike infrastructure does not have a definitively positive or negative effect on business performance. Instead, there are a multitude of other factors outside of planners' control that determine sales or the likelihood that a business closes. There is some evidence to support planning for new bike facilities on low-volume roads, but planners should carefully assess potential impacts when placing lanes on other roadway types. This research can inform future conversations around the relationship between bikes and business, and will hopefully inspire further research into this complex relationship.

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## Introduction

Research has shown that increasing the use of active modes of transportation, including cycling, can have numerous public health and environmental benefits. Providing bicycle infrastructure is one action that cities and counties can undertake to prompt existing cyclists to bicycle more frequently and encourage hesitant would-be cyclists to make trips by bike (Dill and McNeil 2013). One relatively understudied aspect of bicycle infrastructure is its impact on business establishments. Advancing the understanding of this relationship is increasingly relevant as cities across California continue to push for bicycle infrastructure projects. This research explores the bicycle infrastructure and business relationship in two ways. First, secondary business performance data is used to quantify the impact bicycle infrastructure has had on sales and business productivity in two counties in the San Francisco Bay Area. Second, an intercept survey of shoppers on matched pair corridors helps to further illuminate the relationship between bicycle infrastructure, mode choice for shopping trips, and consumer behavior.

Bicycle infrastructure has come under attack throughout the Bay Area from a number of constituencies, including - but not limited to - the business community. Small business owners hold considerable clout when it comes to influencing local policy and decision-making. Few elected officials or government bureaucrats want to find themselves characterized as anti-business, so therefore typically support the case made by local businesses on an individual or organized level (Drennen 2003).

As discussed in more detail below, news articles from different Bay Area sources have revealed a number of merchant justifications for their opposition to bicycle infrastructure, including concerns of parking loss, commercial loading zone obstruction, and/or a perceived 'hassle' for customers arriving by auto. Merchants have opposed bicycle infrastructure installation at all phases of the planning process, including exploratory studies for complete streets plans.

Understanding the relationship between bicycle infrastructure and business outcomes can inform policy in several dimensions. First and foremost, a clear understanding of this relationship, supported by reliable data, will be key in future community outreach for planners at the local, regional, and state levels. Business owners clearly see bicycle infrastructure as a potential threat, but these concerns generally relate to expectations of future harm, as opposed to demonstrated impacts. Equipping planners with facts for outreach could go a long way towards gathering consensus in favor of bicycle infrastructure, which has been shown to produce a number of positive extra-economic effects.

Furthermore, there is considerable heterogeneity within the business community, and all businesses will certainly not be affected by bicycle infrastructure in the same way. Identifying particularly vulnerable industries or business types is a key first step in designing mitigation measures that can help businesses avoid adverse outcomes related to bicycle infrastructure provision. Knowing more about

vulnerable businesses could inform the design and location of bicycle infrastructure, allowing planners to bypass concentrated areas of vulnerable businesses, where possible.

In this report, we examine impacts of three of the four classes of bicycle infrastructure designated by the Federal Highway Administration, Caltrans, and local governments (Caltrans 2017). The primary distinction amongst the different classes is the degree of physical separation between cyclists and vehicular traffic. There are also cost differentials between the different classes.<sup>1</sup>

This report begins with a review of literature on the adoption and performance of bike infrastructure from the perspective of local residents and businesses. After a description of bike facilities – and reaction to them – in the Bay Area, the report discusses the methods used to understand their impact on local businesses. The next section presents descriptive statistics on change in sales, followed by multivariate regression analysis of change in sales and business turnover. The conclusion summarizes findings and offers suggestions for further research.

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<sup>1</sup> Class IV facilities were excluded from analysis because they were not differentiated from Class II facilities in the secondary dataset we used to locate bicycle infrastructure in the Bay Area. This facility type is rare in the Bay Area, though it is becoming increasingly more popular with cyclists and cycling advocates alike, due to the increased perception of safety and comfort provided by the physical separation.

## Background: Bicycle infrastructure and responses in the Bay Area

There are a number of factors that influence where and what type of bicycle infrastructure is installed. From a technical standpoint, motor vehicle traffic volumes and speeds typically dictate the infrastructure class (Caltrans 2017). Higher speeds and volumes tend to translate into more intensive infrastructure projects, with higher-volume streets typically seeing Class II bike lanes or Class IV separated bike lanes. Low-volume and low-speed streets are more appropriate for Class III bike boulevard<sup>2</sup> facilities. Roadway capacity is another determining factor; narrower roads may not have space to accommodate dedicated bike lanes and shared street markers may be the only feasible choice. Class I segregated facilities are only feasible when right-of-way is already in public hands or can be acquired. As such, these facilities are found primarily in parks, along waterfronts and creeks, along highways or rail lines, or on unused rail right-of-way.

However, these decisions are far from being solely on their technical criteria. Stakeholder participation and politics also play a role in deciding if, when, where, and what type of bicycle infrastructure gets installed. Communities throughout the Bay Area have successfully advocated to bring bicycle infrastructure to their neighborhoods, successfully advocated to remove existing infrastructure projects, and successfully fought off proposed infrastructure and studies. When planners and policymakers weigh public input related to bicycle infrastructure projects, business owners are a key constituency.

The following section first describes existing bicycle infrastructure in the Bay Area and then examines how local stakeholders (particularly business owners) have responded to bicycle infrastructure in their communities.

### Types of bicycle infrastructure

Bicycle infrastructure refers to a broad set of amenities that encourage and facilitate cycling as a mode of transportation. This includes facilities for bicycle parking, maintenance, and right-of-way, either in mixed traffic or in exclusive bicycle-only spaces. For the purposes of this report, we focus on right-of-way, or facilities where cycling is encouraged through the provision of infrastructure. There are four classifications of right-of-way bicycle infrastructure in the Bay Area, described briefly below.

#### Class I facilities

Class I facilities (also known as shared-use paths, multi-use paths, or bike paths) are separated from auto traffic. These facilities are for either for cyclists only or are for cyclists and pedestrians; automobile traffic is not permitted. These facilities are

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<sup>2</sup> Bike boulevards are referred to as neighborways and neighborhood greenways in some communities.

typically grade-separated from the roadway and are found parallel to roadways and railways, alongside creeks and drainage ditches, in decommissioned rail right-of-way, and in parks. In urbanized areas, these facilities are most commonly located in parks and are used for both recreation and utilitarian transportation purposes (shopping, commuting, etc.). This type of infrastructure is the least prevalent in the Bay Area; 658 miles of Class I facilities existed in 2014, the most recent year for which this data is available (MTC Regional Bike Facilities 2014).

**Figure 1: Example of Class I Facility: Central County Bikeway, Suisun City**



Image source: [http://www.sta.ca.gov/Content/10088/Biking\\_and\\_Walking\\_in\\_Solano\\_County.html](http://www.sta.ca.gov/Content/10088/Biking_and_Walking_in_Solano_County.html)

### **Class II facilities**

Class II facilities are frequently referred to as “bike lanes.” A Class II facility is defined by a painted line indicating that a certain portion of road space is reserved for people on bikes. Bike lanes are sometimes painted with a solid green treatment to maximize visibility. They can also feature a striped buffer space between the bike lane and passing autos, which increases cyclist comfort by increasing the space between moving vehicles and the bike lane. This is the most prevalent infrastructure type in the Bay Area by lane-miles, with over 1,500 miles of Class II facilities present in 2014 (MTC Regional Bike Facilities 2014).

Figure 2: Example of Class II Facility: Broadway, Oakland



Image source:  
<http://www2.oaklandnet.com/government/o/PWA/o/EC/s/BicycleandPedestrianProgram/OAK043750>

### Class III facilities

Class III facilities are low-speed and low traffic volume roadways meant to be shared by autos and cyclists. Class III facilities are often distinguished into two categories: bike routes and bike boulevards. Bike routes designate a preferred route for cyclists on streets shared with low volumes of automobile traffic, using signage and optional street markings to denote the shared space. These optional road markings are called sharrows, a portmanteau of “share” and “arrow.” Bike boulevards, like bike routes, are also located on low-volume streets. They feature additional traffic calming elements, such as traffic circles, speed humps, pedestrian crossing bulb-outs, and can also feature sharrows to indicate shared space. For the purposes of this analysis, the two sub-types of Class III facilities (bike routes and bike boulevards) are analyzed as one type. The Bay Area was home to just under 1,000 miles of Class III facilities in 2014 (MTC Regional Bike Facilities 2014).

Figure 3: Example of Class III Facility: Milvia Street, Berkeley



Image source:

[https://www.cityofberkeley.info/Public\\_Works/Transportation/Bicycle\\_Boulevard\\_Signage\\_System.aspx](https://www.cityofberkeley.info/Public_Works/Transportation/Bicycle_Boulevard_Signage_System.aspx)

### Class IV facilities

Class IV infrastructure is the newest infrastructure classification permitted by roadway design manuals. Like Class II facilities, Class IV bicycle infrastructure runs parallel to automobile traffic in the same road space. However, like Class I infrastructure, there is an element of physical separation between traveling bicycles and vehicles. Class IV facilities are characterized by vertical physical separation in the form of parked cars, soft-hit posts, planters, or grade separation. This facility type is not distinguished from Class II facilities in the 2014 MTC regional bike infrastructure shapefile, but the region has seen several physically separated facilities installed since the dataset was last updated, including a parking protected bike lane on Telegraph Avenue in downtown Oakland, multiple soft-hit post separated facilities in Berkeley, and soft-hit post separated bikeways on Market Street in San Francisco (shown below).

Figure 4: Example of Class IV Facility: Market Street, San Francisco



Image source: <https://www.sfmta.com/blog/sfmta-public-meetings-may-2-may-16>

Table 1 summarizes the distribution of miles of bike facilities by class for each county in the Bay Area. Each class of bike infrastructure is present in each county. Mileage varies, ranging from over 700 miles of infrastructure (predominantly Class II) in Alameda and Santa Clara Counties to 104 miles of infrastructure in Napa. The distribution of bike facilities varies by county, with some counties providing mostly Class II facilities and others relying on Class III facilities.

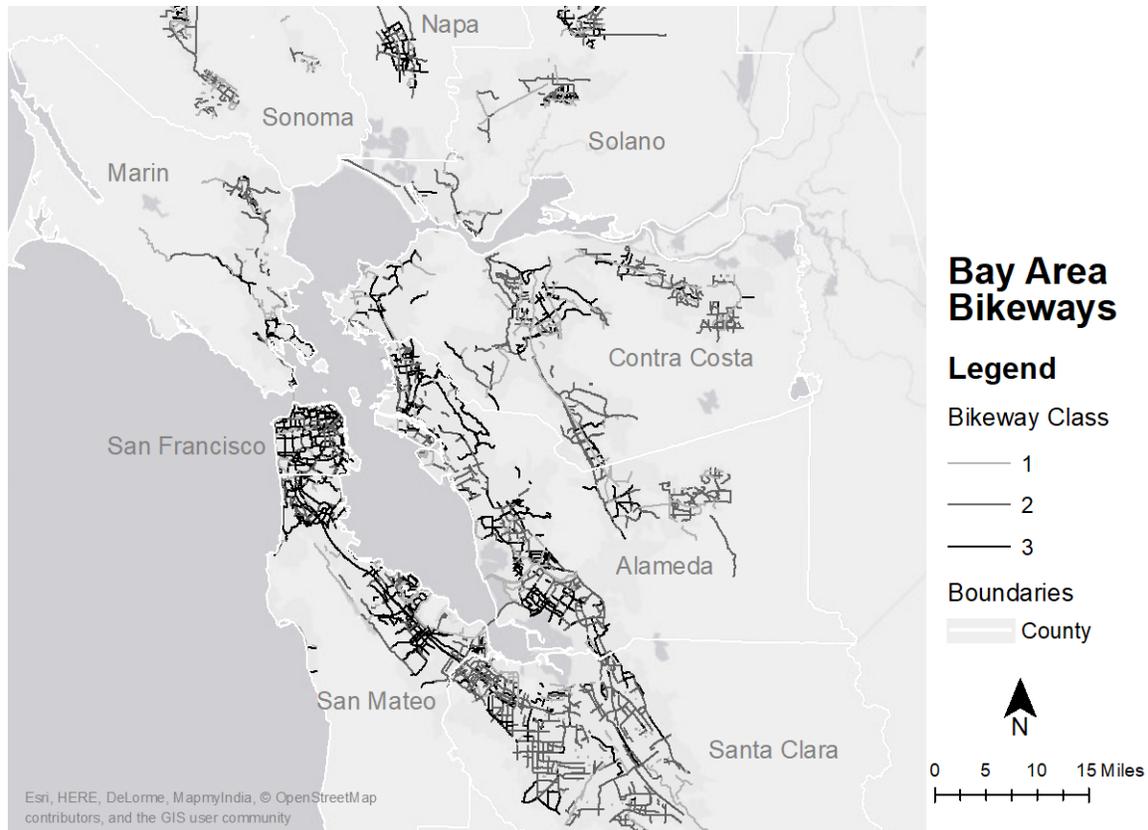
Table 1: Miles of Bicycle Infrastructure by County

County	Class I	Class II	Class III	Total
Alameda	138	335	278	<b>751</b>
Contra Costa	150	263	60	<b>473</b>
Marin	39	60	46	<b>145</b>
Napa	10	57	37	<b>104</b>
San Francisco	32	52	147	<b>231</b>
San Mateo	78	96	241	<b>415</b>
Santa Clara	97	512	117	<b>726</b>
Solano	53	109	51	<b>213</b>
Sonoma	61	115	21	<b>197</b>
<b>Total</b>	<b>658</b>	<b>1599</b>	<b>998</b>	<b>3,255</b>

Data source: MTC Regional Bike Facilities Dataset, 2014

Bicycle infrastructure has been installed throughout the Bay Area in urban, suburban, and rural communities. There are facilities located in isolated areas and predominantly used for recreation, regional connectors that are predominantly used for commuting, and facilities on local streets that could be used for a wide array of trip purposes. Figure 5 illustrates the location of bicycle infrastructure in the Bay Area by facility type. Maps of bicycle infrastructure for each county in the Bay Area are included in Appendix 32.

**Figure 5: Map of Bicycle Infrastructure in the Bay Area**



Data source: MTC Regional Bike Facilities Dataset, 2014

**Local response to bicycle infrastructure in the Bay Area**

Bicycle infrastructure projects are highly political processes throughout the San Francisco Bay Area region. As bicycle-oriented projects are generally considered discretionary, the public is able to exert considerable influence over the end product. Below is a summary of news coverage of contentious bicycle infrastructure projects throughout the Bay Area, which serves as context for understanding the highly-politicized and often heated environment in which this specific type of transportation planning occurs.

A 2011 bike lane on West Spain Street in Sonoma was derailed by strong neighborhood opposition, including a vocal and organized contingent of business owners (Moore 2011). Merchants presented a united front against the project,

submitting a petition signed by every merchant on West Spain Street (Moore 2011). There is currently no bicycle infrastructure on the corridor, and no talk of a future effort to install bicycle infrastructure.

Merchants in San Jose spoke out against a 2014 plan to add four bike lanes to the city's bike network east of downtown (Boone 2014a). A similar plan for installing bike lanes or sharrows on six streets west of downtown was also opposed (Boone 2014b). Although the San Jose Department of Transportation had completed a parking study indicating that the proposed plan would leave adequate parking both east and west of downtown, merchants in the commercial district disagreed (Boone 2014b). A local beauty salon owner was quoted saying, "I'm all for biking, that's a healthy lifestyle I support. But we need to take care of car parking for businesses too" (Boone 2014b). Business owners went on to characterize the removal of parking as a "competitive disadvantage" (Boone 2014b).

Concerned merchants regularly use parking removal as a hot-button issue to stir up opposition. In 2013, an anti-bike lane group by the name of Save Polk Street advocated against a parking protected bike lane in downtown San Francisco (Bialick 2013). The group alerted business owners along the corridor that all parking within a 20 block stretch of the corridor would be removed, contradicting published planning documents by SFMTA (Bialick 2013). Ultimately, three blocks of the corridor received a protected bike lane and the remaining stretch received a temporary pilot treatment that was a much less intensive intervention than originally planned (San Francisco Bicycle Coalition 2017). The planning process for a bike lane on a stretch of Polk Street was reignited in 2017, and time will tell if business owners organize to oppose or support the plan (SFMTA 2017).

Business owners east of downtown San Jose also brought up the issue of commercial loading disruption (Boone 2014a). For businesses that do not have loading docks or loading zones accessible via an alley, curb space for delivery vehicles is an operational necessity. While this oppositional narrative has received less attention and been used less frequently than the complaint of customer on-street parking removal, it is another important element of business opposition.

In Oakland, the parking protected bike lanes on Telegraph Avenue have been touted by Oakland Department of Transportation, as well as transportation publications as an overall success, reducing collisions and increasing bike traffic (Curry 2017). However, that narrative is not universal. Business owners along the corridor doubt that cyclists constitute a substantial pool of customers and characterize driving conditions as unsafe and confusing (Curry 2016).

Merchants have not stopped at simple opposition to bicycle infrastructure projects. In some cases, they have organized to oppose studies that could yield findings that encourage bike infrastructure projects. In 2013, business owners on San Pablo and Solano Avenues in Albany opposed a City Council vote to authorize a Caltrans-funded complete streets study (Esper 2013). Planners said that the study would

result in a guiding document and that any bikeway plan would have to pass through the standard approval process (Esper 2013). However, merchants still opposed the project, seizing on a proposed intersection treatment that would relocate a bus stop and result in the removal of two to three parking spaces (Esper 2013). The campaign against bicycle infrastructure on these corridors was successful, and currently neither corridor includes any form of bicycle infrastructure.

## Literature Review

Policymakers and activists have characterized bicycle infrastructure as both a blessing and a curse for business owners. Despite the abundance of conversation on bicycle infrastructure and its effect on financial outcomes for businesses, there is a relative paucity of academic work examining the relationship. Literature that examines the connection between bicycle infrastructure and business performance can be broken down into the following categories:

- 1) Interview-based analysis of opposition
- 2) Surveys on shopper behaviors and perceptions
- 3) Surveys of merchants on perceived effects and support or opposition
- 4) Analysis of secondary data in the form of taxable receipts, tax records, or other third-party data sources

Studies summarized in this literature review come from academics, planning practitioners, and advocates; the perspectives of different types of authors are examined, along with the analysis and interpretation of the results.

### Interview-based analysis of opposition

Researchers have documented opposition to bike lanes from political economy and critical theory perspectives. When it comes to bicycle infrastructure projects, business owners carry a substantial amount of clout that government actors are beholden to. Often, actions taken to hurt small businesses can become electoral handicaps (Drennen 2003). Opposing bicycle infrastructure can be a political asset; such was the case in the 2010 Toronto mayoral election where a candidate won after promising to “[stop] the war on the car” and pledged to remove a contentious bike lane (Siemiatycki, Smith, and Walks 2016).

Wild et al. (2017) reviewed literature documenting contested bike infrastructure to distill narratives in opposition or support of such projects. The authors found that planners tend to view bike infrastructure through a rational, technocratic lens, characterizing infrastructure projects as minor interventions to make cycling more comfortable. By not engaging in a more critical discussion of the social, economic, and political values embedded in cycling and bicycle infrastructure, planners make themselves vulnerable to being blindsided by “bikelash.” Planners saw the bike lanes as neutral interventions outside the realm of political contention (Vreugdenhil 2013, Vreugdenhil and Williams 2013). In a 2011 opinion piece for *New York Mag*, Shaer described bike lanes as being not “simple strips of pavement festooned with green and white paint” but “sponges for a sea of latent cultural and economic anxieties” (Shaer 2011, p. 2). Situating narratives related to bicycle infrastructure within theory helps to set the stage for larger conversations surrounding gentrification and displacement.

Businesses typically characterize their opposition to bike infrastructure not within the sociocultural narratives described above, but in more practical terms. Through their review of published opposition, Wild et al. (2017) determined that merchants oppose bicycle infrastructure on the grounds of parking loss, driver confusion, and

pedestrian safety. In other words, bicycle infrastructure often represents the first time motorists are being asked to cede road space that was once solely theirs (Vreugdenhil and Williams 2013). Understanding the rationale behind business opposition to bicycle infrastructure informs our later analyses on the effect of on-street parking loss.

### Shopper behavior

Published studies of shopper behavior and travel are primarily survey-based. These surveys ask questions about spending, frequency of patronage, and mode choice. As these studies rely on self-reported figures, they are subject to respondent error. Furthermore, most of the survey designs reviewed here are intercept surveys within relatively small geographic areas, casting doubt on the generalizability of the findings beyond the survey location and sample in question.

In perhaps the best-known paper on the economic impacts of bicycle infrastructure, an intercept survey-based examination of various travel mode users in Portland, Oregon showed that, on average, cyclists spent more at certain business types and patronized them more often (Clifton 2012). This research has been widely celebrated and implemented as a policy tool in bicycle advocacy circles (Maus 2010; Campbell 2015; Szczepanski 2013). That being said, this research addressed a limited sampling frame, surveying only the patrons of eating and drinking establishments and 24 hour convenience stores. Restaurants, bars, cafés, and convenience stores do not represent the wide cross section of businesses that could be impacted by bicycle infrastructure. Namely, purchases at these establishments are small, with almost any purchase being possible to carry home on a bicycle without extra equipment, such as a trailer. Given the heterogeneity of most business communities, a study based on patrons of such a narrow subset of businesses does not lend itself well to generalizability.

Other survey-based research on shopper mode choice and spending habits has been conducted by public agencies in the United States. In 2008, the San Francisco County Transportation Authority (SFCTA) conducted a shopper survey on the Columbus Avenue corridor, finding that those who bicycle, ride in a taxi, or use ‘other’ modes of transportation spent more in the corridor than those who drove or those who took transit (“Columbus Avenue Neighborhood Transportation Study” 2010). The following year, SFCTA conducted another study of mode choice and consumer behavior at several sites in downtown San Francisco, finding that while people who drove to downtown San Francisco shopping destinations spent the most money per trip, they visited business less frequently than shoppers who arrived by walking, biking, or taking transit (Bent and Singa 2009). When the average dollar amount spent was multiplied by the average number of visits per month, drivers spent the least on average and walkers spent the most (Bent and Singa 2009).

In New York City’s East Village neighborhood, a 2012 survey study concluded that the total aggregate spending of bicyclists in the neighborhood was more than drivers, bus riders, taxi users, and ‘select bus service’ riders combined (“East Village

Shoppers Study: A Snapshot of Travel and Spending Patterns of Residents and Visitors in the East Village” 2012). The same survey also found that cyclists had the highest per capita spending in the neighborhood (“East Village Shoppers Study: A Snapshot of Travel and Spending Patterns of Residents and Visitors in the East Village” 2012). This high level of cyclist spending was attributed partially to the fact that cyclists visited local businesses more often and also to increased cyclist traffic due to the addition of protected bicycle lanes on First and Second Avenues (“East Village Shoppers Study: A Snapshot of Travel and Spending Patterns of Residents and Visitors in the East Village” 2012). However, a key limitation of this study is that the survey collected only some demographic information on shoppers. Perhaps most notably, the survey did not ask about income. This is a key missing data point, and attempts to characterize shopper behavior with accounting for income present an incomplete picture.

Two studies of shopper behavior in Davis, California examined spending and mode choice decisions for shoppers. The first study, which contrasts the purchasing behavior of shoppers who arrived by car and bike, found that cyclists spend roughly the same amount per purchase as drivers, and spent more cumulatively over a month, though this difference was not statistically significant (Popovich and Handy 2014). A second study, based on the results of the same survey, sought to understand the decision to cycle for shopping purposes. The authors found that shoppers who agreed with the statement “it is convenient to cycle to my final destination” tended to be more likely to cycle, with drivers exhibiting the same effect when driving was perceived as convenient (Popovich and Handy 2015). This finding is intuitive, but could be extrapolated to consider the role cycling infrastructure has in trip comfort. Individual definitions of convenience are somewhat subjective and heterogeneous, but enabling infrastructure could reasonably be expected to be included as at least part of this judgment. While Davis is a smaller community than other cities studied by research covered in this literature review, it is an extreme case study with a high share of cyclists—in other words, an optimal case that illustrates the maximum potential of infrastructure.

Researchers have also used survey data to model impact of bicycle infrastructure on businesses. A 2008 Australian masters thesis used survey data to examine the shopping habits of cycling- and car-borne customers at inner Melbourne shopping strips. In this research, Lee argued that if public auto parking spaces in commercial areas were reallocated to cyclist parking, 3.6 times the current retail spending could be achieved in the area, with cafés, restaurants, and clothing retailers being the primary beneficiaries of such a shift (Lee 2008). Lee's argument was based upon a customer survey conducted in Melbourne, which estimated auto and cyclist spending behavior. Lee then multiplied the average reported cyclist spending by the number of cyclists that could park in a single auto parking space (Lee 2008, p. 39). Next, the author compared the multiplied cyclist spending product to the average spending of an auto driver occupying the same amount of space, producing the 3.6 times value of cyclists (Lee 2008, p. 39). This line of reasoning, while perfectly logical, is fraught with assumption and context-specific dependencies. For

example, Lee assumes that there is an unlimited number of cyclists waiting in the wings to take the place of auto users and also that the mixture of local businesses would remain unchanged with the loss of auto parking. Further, the paper's geographic context is highly specific – “inner suburban Melbourne” has a particular built environment, climate, and demographic mix, making it difficult to generalize results as applicable to other communities (a problem with any small-geography study of this kind) (Lee 2008, p. 8).

### Merchant perspectives

While research based on surveys of shoppers tended to reveal an expectation that bicycle infrastructure would have a positive effect on business performance, the results from studies examining merchants' perceptions are much more mixed.

Interviews with merchants before and after the installation of a segregated bike path in Sydney, Australia showed initial worry over the effect the infrastructure would have (Crane et al. 2016, 20). Merchants specifically named loss of parking access as a negative aspect of the project. When merchants were interviewed after installation, their concerns seemed largely allayed. Some merchants even admitted their fears might have been overblown. Researchers also interviewed three business owners who had moved to the cycleway after construction was complete. These informants viewed the bicycle infrastructure as a positive amenity of their location. This analysis highlights the divergent narratives of business owners before and after bicycle infrastructure implementation, highlighting the importance of information that helps business owners understand what to expect from a project.

A survey of Irish merchants and shoppers conducted in downtown Dublin showed that merchants routinely overestimated the percentage of customers that arrived by car and underestimated the share of customers arriving on foot or by bus (O'Connor et al. 2011). This study highlights misperceptions that may exist within the merchant community, providing a possible explanation for Crane's finding of initial merchant apprehension surrounding bicycle infrastructure projects. A similar parallel survey of restaurateurs and patrons in downtown Brisbane, Australia revealed that 18% of patrons arrived by car as opposed to the 52% estimated by restaurant owners (Yen, Tseng, and Ghafoor 2015). Seven percent of customers arrived by bike, instead of the two percent expected by the business owners. The survey also revealed that restaurateurs underestimated the amount of money spent by users of active modes and transit (Yen, Tseng, and Ghafoor 2015).

A 2003 interview-based study examined bicycle lanes installed in San Francisco's Mission District uncovered strong merchant support for the bikeways, including the belief amongst 65% of interviewees that bicycle lanes have had positive impacts on sales and their business overall (Drennen 2003). This research was published 15 years ago, when dynamics in the Mission were quite different from those existing today.

There are several studies that investigate the effect of bicycle infrastructure through self-reported data on sales and profit collected through surveys of business owners. Planners in Vancouver, British Columbia surveyed business owners after the installation of a separated bikeway. The sampling frame consisted of all ground floor businesses abutting the bikeway, as opposed to previous studies which focused on a subset of businesses. Business owners reported declines in sales and profit. The report predicted that “the moderate negative impact of the lanes will diminish over time as long as mitigation strategies take effect” (Stantec Consulting Ltd. 2011, p. iii).

Dual survey-based studies performed on the Bloor Street corridor in Toronto found that overall, merchants and shoppers alike supported the addition of bicycle lanes to the right-of-way and that those arriving by bicycle, transit, and foot were likely to spend more in the commercial district than those arriving by auto (Sztabinski 2009 and Forkes and Smith-Lea 2010). Survey results from the 2009 research revealed that “75% of merchants thought a bike lane or widened sidewalk would improve or have no effect on business, and patrons preferred a bike lane to widened sidewalks at a ratio of almost four to one” (Sztabinski 2009, p. 23). The 2009 analysis roundly concluded “that merchants in this area are unlikely to be negatively affected by reallocating on-street parking space to a bicycle lane. On the contrary, this change will likely increase commercial activity” (Sztabinski 2009, p. 1). In 2010, a similar survey-based analysis conducted in the area found that more than half of merchants believed reducing parking and adding a bike lane would increase or have no impact on their number of customers” (Forkes and Smith-Lea 2010, p. 10). The same study found that “The majority of people surveyed (58%) preferred to see street use reallocated for widened sidewalks or a bike lane, even if on-street parking were reduced by 50%” (Forkes and Smith-Lea 2010).

When interpreting the results of any survey-based research, it is important to remember that responses are self-reported and thus vulnerable to misremembering, exaggeration, and other forms of bias. These analyses – as with most of the research cited in this literature review– do not cite sales tax receipts or other business sales data, however, and measure only the perception of merchants and customers, not hard sales data or measures of productivity. Although perception is important, analysis using unbiased and universal data could provide a clearer picture of the effect of bicycle infrastructure on business performance. Tax and sales data is relatively hard to come by, and often comes from proprietary sources that charge hefty access fees. The dearth of published research using these sources is understandable.

### **Analysis of secondary data**

Analysis of secondary data to discern impacts of bicycle infrastructure on business outcomes has generally used taxable receipt data to gather an unbiased picture of trends in business performance before and after the implementation of bicycle infrastructure. However, such analysis has often been conducted in an opaque way on datasets that are not publicly available for re-analysis. As this paper relies

primarily on analysis of secondary business performance data, these sources have been of critical importance to the development of our methodology.

A 2012 New York City Department of Transportation (DOT) publication claimed increases in retail sales of “up to 49%” along new protected bicycle lanes on Ninth Avenue from 23<sup>rd</sup> to 31<sup>st</sup> streets in Manhattan, compared to increases of three percent borough-wide (“Measuring the Street: New Metrics for 21st Century Streets” 2012, p. 4). Similarly, a 2017 Oakland DOT planning document using sales tax receipt data stated that retail sales in the Telegraph corridor had increased nine percent year-over-year after a protected bicycle lane intervention (Fine 2017). However, in neither DOT study were the bicycle lane corridors compared to control corridors or nearby areas, making it difficult to separate the effect of bicycle infrastructure from general economic trends or other neighborhood-wide changes

A taxable retail sales-based analysis of two bicycle lane interventions in Seattle, Washington included dual control regions for each of the two study areas, showing no negative sales impact on businesses resulting from either bicycle lane intervention (Rowe 2013). The study suggested the possibility of a wildly successful economic impact produced by a climbing lane installed on NE 65<sup>th</sup> Street, which removed 12 parking spots and was correlated with a 400% increase in sales in the district (Rowe 2013, p. 2). Rowe 2013 used a more sound and transparent methodology than both the NYC and Oakland DOT reports, with more appropriate control regions and greater discussion of the results.

A 2018 study of three bicycle infrastructure implementations in San Francisco found used National Establishment Time Series (NETS) sales data to examine business outcomes in terms of sales per establishment and sales per employee, finding that results were mixed but rarely negative (Poirier, 2018). The primary findings of this research effort were that different types of businesses (specifically, local-serving vs. non-local serving) experience bicycle infrastructure implementations differently and that the bicycle advocate mantra of ‘bike lanes are always good for business’ is likely not true.

### **Bicycle infrastructure and commercial gentrification**

A cross-cutting theme across the research is the relationship between bicycle infrastructure and commercial gentrification, a theme associated with the broader literature on environmental or “green gentrification” (Gould and Lewis 2017). This discussion connects public and private investment in environmental amenities with a subsequent increase in real estate prices, gentrification, and displacement. Municipalities often promote livability as a tool to attract capital and its talent (Kreugeran Gibbs 2007, Raco 2005). For instance, in Seattle, an ecological agenda for urban open spaces displaces and excludes the homeless (Dooling 2009). In Vancouver, developers use discourses of sustainability for marketing purposes and to spur gentrification (Quastel 2009).

The Complete Streets movement itself raises questions. Complete Streets processes revamp streets, with the idea of providing safe access for all users, instead of simply facilitating automobility (McCann and Rynne 2010). Yet, as Zavestoski and Agyeman ask, ‘complete for whom?’ Most complete street projects are implemented on just a few blocks, with designs that try to accommodate diverse users, including disabled pedestrians, bicycles, and delivery trucks, but tend to fall short in meeting some needs (Zavestoski and Agyeman 2014).

The primary method for examining the relationship between bike infrastructure and gentrification has been through interviews and ethnography. Valencia Street in San Francisco has served as a case study site for several critical theory articles. The early advocacy of the San Francisco Bike Coalition adopted a narrative of the project centered on the uptick in commercial activity but ignoring the social and cultural ramifications of the project (Stehlin 2015). The corridor is described as a “key material, ideological, and practical linkage between bicycle infrastructure, cosmopolitan urbanity, and economic growth” (Stehlin 2015, p. 125). This critique of the progressive concept of urban livability for its disregard of gentrification is shared by Rankin and McLean, who point to the correlation between bicycle infrastructure and other amenities designed to appeal to creative types and initiate economic change in urban commercial corridors (Rankin and McLean 2015). Lubitow further unpacks this relationship through a study that used interviews and observation to uncover the dynamics behind community opposition to a bike lane on Paseo Boricua in Chicago, a principal artery for the city’s largest Puerto Rican neighborhood (Lubitow, Zinschlag, and Rochester 2016). Some community members revealed that they were not opposed to bicycle infrastructure itself, but rather the top-down approach that the city had taken to the planning process. Others identified the project as “pav[ing] the way for gentrification” (Lubitow, Zinschlag, and Rochester 2016, p. 2643). Since the research summarized in this section of the literature review relies primarily on small sample interviews and review of historical documents, there are limitations to the generalizability of the findings.

### **Role of advocacy publications**

Much of the findings summarized in this literature review have been interpreted and reported by cycling advocacy groups in an attempt to convince business owners that bicycle lanes increase sales (Tolley 2011, Flusche 2012, Szczepanski 2013, Andersen and Hall 2014). In an attempt to advocate across broad geographies, advocates have frequently taken corridor-specific studies and cast them as generalizable findings, a practice that could be interpreted as misleading. The need for analysis that speaks to the effect of bicycle infrastructure across a broad geography is evident, though difficult to meet, given the lack of comprehensive data sources reporting business performance metrics or the location of bicycle infrastructure. This is a gap in the literature that this study seeks to fill through analysis at the county and regional levels.

## Summary

Researchers from across disciplines have examined the impact of bicycle infrastructure on business performance, but a clear narrative has not emerged. While advocacy organizations have seized on research that finds a positive effect on businesses, there are limitations to applying this research at the broad scale that advocates would like. Research has also indicated that there could be a negative impact for businesses on corridors with bike infrastructure. Merchant opposition has been clearly documented, but research that looks at merchant attitudes over time tends to show a more neutral or even pro-bike infrastructure stance. This research paper seeks to carve a niche in the literature by applying empirical methods to analyze the effect of bicycle infrastructure on businesses at the county level, as opposed to simply examining conditions at the corridor level, as the majority of studies have done. The survey of shoppers on matched pair corridors complements this research by examining the effect of such infrastructure on an individual's purchase decisions. Through the triangulation of these multiple methods, we hope to paint a more comprehensive picture of the state of bicycle infrastructure and business performance in the Bay Area.

## Methodology

### Data sources

#### Business performance

Business performance data come from the National Establishment Time Series (NETS) database, a proprietary database assembled by Don Walls & Associates that combines Dun & Bradstreet data on individual establishments into an annual time series from 1990 through 2014. As is typical of business data, this database has shortcomings, including inaccuracy of data (both self-reported and estimated) and infrequency of updates (Kroll, Lee, and Shams 2010; Neumark, Zhang, and Wall 2007). NETS is a census of businesses with 99% of businesses in the state of California reporting (Walls & Associates 2017). Businesses of all sizes are included in the dataset, from incorporated freelancers earning a few hundred dollars per year to healthcare systems earning millions of dollars and employing thousands. The dataset includes data about businesses that were open at any time between 1990 to 2014, inclusive.

For the purposes of our analysis, the primary NETS variables of interest were sales, number of employees, physical address, industry (six-digit NAICS code), first year, and last year. Other variables, including the number of related businesses, were also used as control variables in our models.

#### Roadway characteristics

We hypothesized that business performance would vary based on corridor characteristics. Measuring the performance of businesses located on principal arteries against that of businesses on side streets is not an intuitive apples-to-orange comparison. As such, we created a gradient-based corridor typology, with four corridor types ranging from statewide connectors to neighborhood streets.

Corridor data came from OpenStreetMap (OSM), a global open-source mapping database. This OSM data includes linear features representing roads and paths. Each linear feature has a road class designation, with 27 total designations present in the dataset (Ramm 2017). Feature classes that represented horse trails, recreational bike facilities, hiking trails, staircases, ferry routes, private roads, or features of unknown uses were excluded from the corridor typologies. Interstates were also excluded from the typologies classification due to their limited access design.<sup>3</sup> The remaining feature classes were condensed into four corridor types, summarized in Table 2.

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<sup>3</sup> Bicycles are rarely permitted on freeways, although there is a brief stretch of I-580 in Richmond, CA that permits bicycles.

**Table 2: Miles of roadway types by county**

<b>County</b>	<b>Road classification</b>	<b>Miles</b>
Alameda	Primary roads	1,990
	Secondary roads	3,348
	Tertiary roads	4,073
	Neighborhood roads	11,170
<b>Total</b>		<b>20,581</b>
San Francisco	Primary roads	1,176
	Secondary roads	1,716
	Tertiary roads	3,150
	Neighborhood roads	7,899
<b>Total</b>		<b>13,941</b>

Primary roads are often state-owned and -operated facilities that promote statewide connectivity. This type also includes roads designated as trunk corridors, which are roadways similar to primary roads that also feature a median dividing traffic flows in opposite directions. In the Bay Area, these corridors include the East Bay’s University Avenue or Broadway. While these facilities are typically among the corridors with the highest auto traffic volumes, some are equipped with bike infrastructure. Bike infrastructure on primary roads is predominantly Class II, due in part to the need for clear delineation of road space for bicycles due to high traffic speeds and volumes.

Secondary roads offer regional connectivity. These facilities are also home to high volumes of auto traffic, and their bike infrastructure tends to consist of mostly Class II facilities as a result. In the Bay Area, roads classified with the corridor type of secondary roads include Mission Street in San Francisco and Telegraph Avenue in Oakland.

Tertiary roads provide local connectivity. This is OSM’s lowest-volume “major road” classification (Ramm 2017). In the Bay Area, these corridors typically feature a mix of Class II and Class III facilities. Examples of tertiary roads in the Bay Area include Valencia Street and 24<sup>th</sup> Street in San Francisco or College Avenue in Berkeley.

The fourth and final road type, neighborhood streets, is by far the largest in terms of mileage. This type includes roads designated as neighborhood streets, service facilities (alleys and other narrow access streets), and living streets, or streets where pedestrian access is privileged. Due to the relatively low traffic volumes and speeds, these corridors tend to be designated as bicycle boulevards and feature Class III facilities, although some are equipped with Class II bike lanes.

### Bicycle infrastructure

Data on the location and class of bicycle infrastructure came from two sources. The Metropolitan Transportation Commission (now Bay Area Metro), the Metropolitan Planning Organization that oversees regional planning in the Bay Area, publishes data on the location and class of bike infrastructure throughout the region. This dataset features information on bicycle infrastructure through the year 2014, which conveniently matched the time frame of the NETS dataset. The MTC data was available in spatial and tabular formats. A drawback to the MTC data is that it does not include installation date for any infrastructure. However, installation dates were manually assigned to facilities in Alameda County using historic satellite imagery from Google Earth.

The MTC dataset was supplemented by San Francisco Municipal Transportation Agency (SFMTA)'s Bikeway Network dataset. This dataset, also in dual file format, includes the location and class of bicycle infrastructure for San Francisco, as well as installation month and year.

### On-street parking

On-street parking counts before and after the installation of bicycle lanes were conducted for San Francisco. Parking counts were only conducted for corridors with Class II facilities, as Class III bikeways do not affect the number of parking spaces and Class I facilities were not found in San Francisco outside, of parks. Parking counts were not conducted for Alameda County due to resource constraints, though future research should add this data.

On-street parking counts were conducted using Google Street View. Google provides Street View imagery from 2007 to 2018, though not all locations have imagery dating back to 2007. Researchers counted the number of parking spaces on a given block in the most recent Google Street View imagery, then navigated back in time to Street View imagery from before the bike lane was installed and counted the number of on-street spaces present before installation.

Researchers counted the number of unpainted curb parking spaces as well as the number of yellow, green, white, and blue spaces. Yellow curb space is meant to facilitate commercial loading by designating space specifically for commercial vehicles. Knowing the number of yellow loading zone spaces was of special interest, given that difficulty with commercial loading is a concern voiced by merchants who oppose bicycle infrastructure projects. Noting the placement of white curb space for passenger loading is also potentially of interest, as e-hailing companies and taxis make use of these spaces to drop off passengers. As e-hailing companies are a new phenomenon, there has not been as much of a public clamor for passenger loading space as there has been for commercial loading space, though this is a growing issue in San Francisco and elsewhere.

The change in the number of parking spaces of each type (standard, color curb) was then calculated using the before and after counts. There were instances where the

bike lane predated Street View; in those cases, parking change was not calculated. This parking change variable is used in the sales and business turnover models for San Francisco.

It is possible that parking spaces were removed on corridors without bicycle infrastructure for projects other than bicycle infrastructure, such as pedestrian facilities, parklets, bike corrals, or transit facilities. Future research should capture a more comprehensive picture of change in on-street parking.

## Data processing

### NETS data cleaning

As mentioned in the dataset description, NETS covers all businesses in the state of California. We first limited the data to businesses in Alameda and San Francisco Counties using a variable in the NETS dataset indicating FIPS code. Next, we geolocated the businesses using latitude and longitude from NETS to confirm that the address was within the county corresponding to the designated FIPS code. Businesses that had been assigned an incorrect FIPS code were eliminated.

The NETS dataset also noted the years in which businesses moved. For businesses that did move, NETS provides the establishment’s first and last address. The dataset did not provide information on intermediate addresses for businesses that moved more than once during their lifespan. As such, businesses that moved more than once were excluded from analysis. This represented a small fraction of businesses in the sample, at roughly 2,500 businesses in the Bay Area.

Once we were confident that the dataset consisted of only businesses that were actually located in Alameda and San Francisco counties, the dataset was subset again to include only storefront-type businesses. Using six-digit NAICS codes, we selected a list of 106 industry designations that characterized businesses of interest – storefront retail, food service, and other service-providing businesses that stand to be affected by bicycle infrastructure. Businesses that aren’t dependent on consumer access, such as office-based workplaces or manufacturing sites, were excluded from analysis. The list of storefront-type NAICS codes can be found in Appendix 1.

To analyze trends along industry lines, 10 industry classifications were created using six-digit NAICS codes. Not every business in the sample falls into one of the industry categories, but they do represent prominent industries. Table 3 lists the ten industry classifications and presents the total number of businesses within each classification. The industry breakdowns are generally similar between the two counties. Note that percentages do not sum to 100 due to rounding. The NAICS codes that correspond to each industry grouping are found in Appendix 2.

**Table 3: Industry by number of businesses**

Industry	San Francisco County		Alameda County	
	N	%	N	%

Bar	1,177	2%	765	1%
Restaurant	7,714	13%	7,269	9%
Grocery	3,497	6%	4,000	5%
Personal goods	7,708	13%	10,457	13%
Home goods	4,021	7%	6,664	9%
Services	7,952	14%	10,300	13%
Entertainment	61	<1%	68	<1%
Financial	2,782	5%	3,552	5%
Health	10,349	18%	12,864	16%
Automobile-oriented	2,585	4%	8,882	11%
Uncategorized	9,652	17%	13,480	17%
<b>Total</b>	<b>57,498</b>	<b>99%</b>	<b>78,301</b>	<b>99%</b>

A major component of this research is analysis of change over time. All sales figures were inflation-adjusted to 2014 dollars using CPI rates from the Bureau of Labor Statistics.

The final step in NETS data cleaning was eliminating outlier businesses. Businesses with one or zero employees were eliminated, as were businesses with sales larger than the 99<sup>th</sup> percentile for that given year.

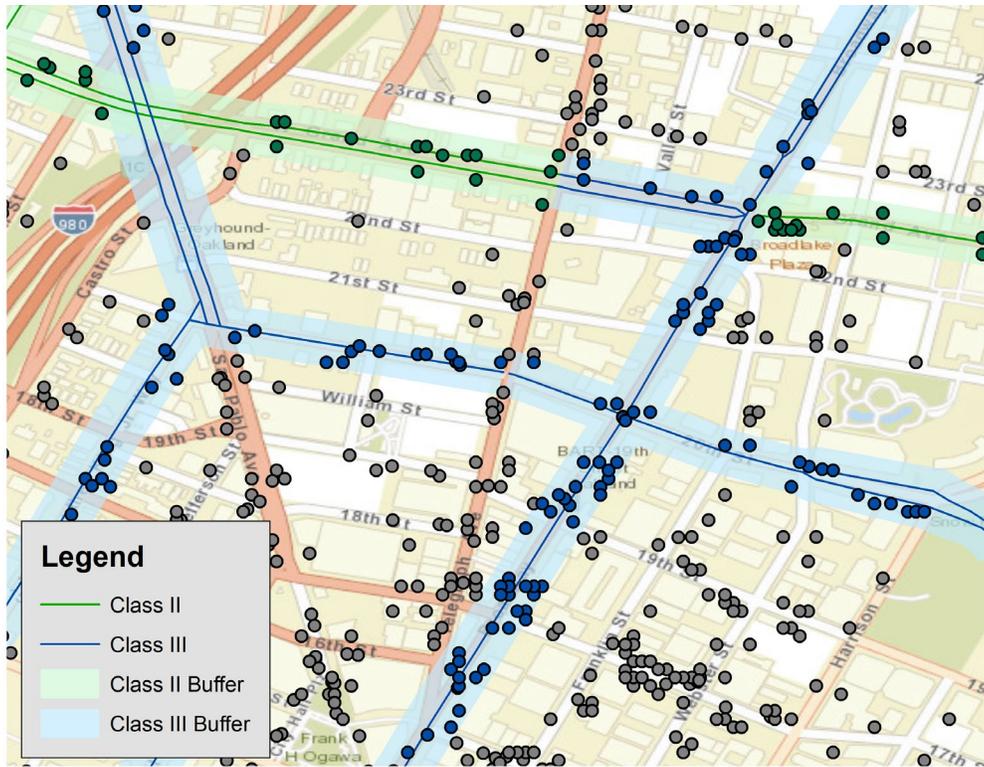
#### Designating abutting businesses

Throughout this report, businesses that abut bicycle infrastructure will be compared against businesses that do not abut bicycle infrastructure. The following process was used to assign an abutting dummy variable to businesses that directly face bicycle infrastructure. The bicycle infrastructure shapefiles from MTC and SFMTA both used street centerlines to represent bicycle infrastructure. Meanwhile, NETS data placed the point feature for each business inside parcels adjacent to streets.

Using ArcGIS, we drew a 100-foot buffer around the bicycle infrastructure street centerlines to capture the nearby business point features. Buffers of varying lengths were spot-tested to determine the minimum distance necessary to capture abutting businesses without also beginning to capture non-abutting businesses within the buffer zone. The polygon buffer features could then be spatially joined to business point features to create a dummy variable for abutting businesses. This process was repeated for Class I, Class II, and Class III corridors to create dummy variables for each type of infrastructure. Figure 1 illustrates the process. It depicts corridors with Class II facilities and their buffers in green and Class III corridors and their buffers in blue. Businesses that fall within the green or blue buffers would be coded as being Class II or Class III, respectively. All other businesses would be coded as non-abutting. Businesses at the intersection of two or three buffers were dropped from analysis, as it is challenging to reliably determine whether the business had an entryway on the Class II or Class III corridor. However, these businesses represented a small fraction of all businesses in the sample, at less than 1% of

businesses in Alameda County (315 of 78,301) and less than 3% of businesses in San Francisco (1,489 of 57,498).

**Figure 6: Bike infrastructure buffer process in downtown Oakland**



**Other built environment variables**

Population density classes were calculated using percentiles, with block groups under the 33<sup>rd</sup> percentile being classified as rural, block groups in the middle tercile classified as suburban, and block groups in the 66<sup>th</sup> percentile and above classified as urban. This translated to cutoff points of seven people per acre or less for rural block groups, eight-15 people per acre for suburban block groups, and more than 15 people per acre for urban block groups.

To compare businesses across corridors that see similar traffic volumes and offer similar degrees of local, regional, or statewide connectivity, we used our four-tiered roadway classification (see Table 2). The four road classifications are inclusive, meaning each business in the sample falls into one of the four classifications. Table 4 summarizes the number of businesses located on each of the four roadway classifications, as well as the number of businesses abutting and not abutting bike infrastructure.

**Table 4: Abutting and non-abutting businesses by roadway classification**

County	Roadway classification	Bicycle infrastructure	Businesses	
			N	Percent
Alameda	Primary roads	Non-abutting	8,632	91%

		Abutting	808	9%
	Secondary roads	Non-abutting	11,793	96%
		Abutting	459	4%
	Tertiary roads	Non-abutting	9,652	94%
		Abutting	585	6%
	Neighborhood roads	Non-abutting	43,263	96%
		Abutting	1,828	4%
<b>Total</b>			<b>77,020</b>	
San Francisco	Primary roads	Non-abutting	5,405	65%
		Abutting	2,952	35%
	Secondary roads	Non-abutting	7,012	74%
		Abutting	2,401	26%
	Tertiary roads	Non-abutting	7,190	51%
		Abutting	6,942	49%
	Neighborhood roads	Non-abutting	19,066	78%
		Abutting	5,389	22%
<b>Total</b>			<b>56,375</b>	

Alameda County shows relatively little variation in the share of businesses abutting bike infrastructure, with an average hovering between four and nine percent of all storefront businesses abutting corridors of any class. As a more urban county, San Francisco shows a considerably higher proportion of businesses abutting bike infrastructure, with upwards of a 50/50 split of businesses on tertiary roads abutting bicycle infrastructure.

# Modeling the relationship between bicycle infrastructure and business performance

## Introduction

The following section presents several statistical models that illuminate the relationship between bicycle infrastructure and business performance. The models control for business, neighborhood, and corridor characteristics to present the effect of bicycle infrastructure in isolation.

## Dependent variables

There are two primary outcomes of interest in determining the relationship between bicycle infrastructure and business performance: change to sales and likelihood of business turnover. The sales change model looks at the effect of bicycle infrastructure on the change to sales over the years 1990 to 2014. The change in average sales is calculated as the average sales for a business in the years prior to the installation of bike infrastructure subtracted from the average sales for a business after the installation of bike infrastructure, reported in 2014 dollars. For non-abutting businesses, the same change is calculated using the year of installation for the closest bike infrastructure project as the year around which before and after averages are calculated. Examining the change in total sales outcome is of interest to business owners and planners alike because sales is perhaps the most easily understood measure of business performance. Given that different infrastructure types may have different effects, models are presented for each class of bicycle infrastructure in each county.

The second set of models differs from the first set in that it does not model the linear relationship between bicycle infrastructure and business performance. Rather, these models estimate the probability of a business closing versus staying open. Understanding whether location on a corridor with bicycle infrastructure changes the likelihood that a business will close is of interest because remaining open is one of the most basic definitions of business success. This section also includes models for each infrastructure type in each county.

The San Francisco models include a set of variables related to parking and curb management, including the change in number of on-street parking spaces and dummy variables indicating the presence of commercial or passenger loading zones on the block where the business is located.

There were very few businesses abutting Class I infrastructure (zero businesses for San Francisco and 17 businesses for Alameda County). This sample size is so small that estimating the effect of location on Class I infrastructure would not be statistically robust. As such, this section only presents models for Class II and Class III infrastructure.

## **Independent variables**

### **Bicycle infrastructure**

Bicycle infrastructure variables are the primary variables of interest. The model accounts for infrastructure class with a dummy variable indicating whether or not the business abuts a Class I, Class II, or Class III corridor. This enables us to tease out the effects that different types of infrastructure may have.

### **Corridor characteristics**

The second set of independent variables relates to characteristics of the corridor. Because comparing businesses on a principal artery to businesses on a side street is not necessarily a fair comparison, the model includes the four ordinal roadway classifications. Neighborhood streets are excluded from the model and serve as the reference category against which the other three road classes can be compared.

### **Business characteristics**

There are a number of business characteristics that could theoretically influence business performance outcomes. The model takes these into consideration and calculates their effect apart from the effect of the bicycle infrastructure. The major distinction taken into account is the industry (one of 10 overarching industry categories).

Other variables related to the business include business age a dummy variable representing whether or not the business was a chain. The chain dummy variable includes both national and local chains, as any business that had one or more related locations in the dataset was marked as a chain.

### **Neighborhood characteristics**

The neighborhood where a business is located can also have an effect on business performance. Understanding and controlling for the built environment characteristics and demographic profile of the neighborhood was a key step in the modeling process. All neighborhood characteristic data points are aggregated at the census tract level, which is the most granular level at which data from the US Census Bureau's American Community Survey are available for the years 2009 to 2014. Neighborhood characteristic data are current to the year of installation (or the year of installation for the nearest bicycle infrastructure project, in the case of non-abutting businesses). ACS data were not available for the years before 2009. In these cases, data from either the 2000 Census or the 2009 ACS were assigned to the business. Businesses on bike infrastructure projects that were installed before 2004 were assigned demographic data from the 2000 Census; businesses on infrastructure installed between 2004 and 2009 were assigned demographic data from the 2009 ACS. There were very few businesses abutting bike infrastructure installed in the 1990s. Data from the 1990 Census were not available online, so they could not be used for model estimation.

To understand the urban form surrounding the businesses in the sample, we looked at density of people and businesses. The model includes an establishment density variable calculated as number of businesses per acre and dummy variables representing urban, suburban, and rural densities. The density dummy variables were calculated using terciles: the first tercile was marked as rural (density of fewer than seven residents per acre), the middle tercile was marked as suburban (seven to 15 residents per acre), and the third tercile was marked urban (more than 15 residents per acre). The final built environment variable represents the percent of housing that is renter-occupied.

To better understand the profile of people living in the neighborhood surrounding businesses, the model also controls for various economic and demographic features. Median<sup>4</sup> household income in 2014 dollars represents the general affluence of the surrounding neighborhood. The model also includes some information on the racial and ethnic composition of the neighborhood (with variables that account for the percent of the population that identifies as Hispanic or of Spanish origin, and the percent of the population that is African-American).

#### **Parking and curb management (San Francisco Class II corridors only)**

Google Street View was used to manually count the change in on-street parking before and after the installation of bike lanes. Counts were only registered on corridors where Class II facilities were installed after the year 2007. Because Google Street View imagery first became available in 2007, bike lanes installed before 2007 could not be considered using the Google Street View methodology. Only corridors with bike lanes were evaluated for on-street parking change because Class III facilities do not typically necessitate the removal of on-street parking and Class I facilities in San Francisco are largely located in parks without on-street parking.

Additionally, this category includes dummy variables for the presence of passenger loading zones (white curb) or commercial loading zones (yellow curb) on corridors with bike infrastructure. This data was collected during the Google Street View parking counts, so it is only available for corridors with bike lanes installed after 2007. Curb management is an emerging issue of great interest to planners and businesses, especially given the growing role that e-hailing companies, such as Uber and Lyft, play in the transportation system. The introduction of this report summarizes news articles where business owners cite loss of loading access as a chief concern related to bicycle infrastructure projects. Future research should account for all loading zones in San Francisco or the Bay Area to produce a more comprehensive assessment.

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<sup>4</sup> In a region with growing income inequality, the median controls for unusually high incomes and presents a more representative picture of incomes in the neighborhood.

## Interactions

The models also include two sets of interactions: one between roadway class and bike infrastructure and one between business type and bike infrastructure. Including interaction terms within the model helps identify and quantify any differing effects that bike infrastructure has based on the traffic volume of the corridor and the business industry.

The interaction between bike infrastructure and corridor type seeks to understand if bike infrastructure has a different effect when installed on high-volume primary or secondary roads, as opposed to low volume tertiary or neighborhood roads. The interaction between bike infrastructure and business type has a similar aim: to identify whether certain industries benefit from bike infrastructure and others are negatively affected.

## Sales change models

### Average sales change

Table 5 summarizes the overall change in sales for businesses in San Francisco and Alameda Counties, as well as the change in sales for non-abutting businesses and businesses abutting Class II and Class III corridors. Sales declined across the board, likely due to an overall market shift away from storefront retail towards e-commerce. Sales in Alameda County declined more across all subsets of the data except for businesses abutting Class II infrastructure. Businesses on Class II bike infrastructure saw the largest declines in both counties. The smallest decline was seen by businesses abutting Class III infrastructure in San Francisco and non-abutting businesses in Alameda County, though the difference between non-abutting and Class III facilities in both counties was small. These patterns could be due to an effect of bike infrastructure, or they could be due to other patterns related to the types of businesses located on corridors with bike infrastructure, the corridors where bike infrastructure is located, or the neighborhoods in which the businesses are located. The following models seek to control for other factors that could have an effect on how businesses fare in terms of sales to isolate the relationship between bike infrastructure and change in sales.

Table 5: Average change in sales (all years)

	San Francisco County	Alameda County
Average sales change	-\$49,967	-\$53,223
Average sales change non-abutting businesses	-\$47,676	-\$53,519

Average sales change businesses abutting Class II	-\$75,342	-\$64,524
Average sales change businesses abutting Class III	-\$40,239	-\$54,596

### San Francisco County

Change in sales volume is perhaps the most easily interpretable definition of business performance. An increase in sales volume suggests a business is doing well, while declining sales could indicate the business is headed for closure. Table 6 presents the results of a linear regression that models the change in sales before and after the installation of Class II facilities in San Francisco. The change in sales for abutting businesses is calculated as the average of sales across all years 1990-2014 before the installation of the bicycle infrastructure, subtracted from the average of sales across all years after the installation of infrastructure. The installation year itself is excluded from analysis. Non-abutting businesses were assigned the same installation date as the nearest corridor with bicycle infrastructure using a nearest neighbor spatial join in ArcGIS. For simplicity, the nearest corridor was calculated using Euclidian distance instead of network distance. The average change was then calculated in the same manner as for abutting businesses.

**Table 6: Change in sales for Class II facilities in San Francisco County**

	Coefficient	Sig.	Standard Error	T Statistic
<b>Roadway characteristics</b>				
Class II	2,642		18,200.00	0.15
Primary road	-18,780	*	9,756.95	-1.93
Secondary road	-31,630	***	10,600.00	-2.98
Tertiary road	-165		8,056.16	-0.02
Change in on-street parking	194		2,305.05	0.08
Commercial loading zone	53,200		60,600.00	0.88
Passenger loading zone	4,686		28,100.00	0.17
<b>Business characteristics</b>				
Years Old	-2,094	***	202.51	-10.34
Chain	-18,280		12,200.00	-1.50
Bar	57,940	***	17,600.00	3.28
Restaurant	15,720		10,600.00	1.48
Grocery	1,400		13,800.00	0.10
Personal goods	287		11,000.00	0.03
Home goods	-10,250		14,900.00	-0.69
Services	40,050	***	10,200.00	3.92
Entertainment	39,860		68,600.00	0.58

Financial services	42,630	**	19,400.00	2.20
Health services	-12,660		9,989.91	-1.27
Automobile-oriented	-26,520		17,200.00	-1.54

#### Bike infrastructure and road class interactions

Primary roads x Class II	10,470		21,100.00	0.50
Secondary roads x Class II	-17,200		24,600.00	-0.70
Tertiary roads x Class II	-19,690		16,500.00	-1.20
Neighborhood roads x Class II	29,060	**	14,300.00	2.04

#### Business and bike interactions

Bike x bar	-16,740		60,400.00	-0.28
Bike x restaurant	-25,580		31,500.00	-0.81
Bike x grocery	-10,440		34,900.00	-0.30
Bike x personal goods	-956		35,400.00	-0.03
Bike x home goods	-94,250	**	43,000.00	-2.19
Bike x services	-17,080		30,400.00	-0.56
Bike x entertainment	0		0.00	1.64
Bike x financial services	0		0.00	-1.35
Bike x health services	-14,500		29,200.00	-0.50
Bike x automobile-oriented	-162,800	***	41,500.00	-3.92

#### Neighborhood characteristics

Establishment density (businesses/ per acre)	-414		451.26	-0.92
Urban	16,040		10,200.00	1.58
Suburban	0		0.00	1.29
Median household income (\$1,000)	-212		158.19	-1.34
Percent housing renter occupied	40		230.24	0.17
Percent Latino	199		263.64	0.76
Percent Black	-279		496.89	-0.56

#### Installation year

1996	-5,685		21,600.00	-0.26
1997	2,892		15,000.00	0.19
1998	-4,023		48,700.00	-0.08
1999	-29,730		23,700.00	-1.26
2000	4,936		21,200.00	0.23
2002	2,532		21,900.00	0.12
2003	-13,240		15,000.00	-0.89
2004	35,090		24,200.00	1.45
2005	-2,581		9,595.55	-0.27

2006	-14,390		11,500.00	-1.25
2008	6,881		38,800.00	0.18
2009	22,660	*	13,100.00	1.72
2011	21,860	**	10,300.00	2.11
2012	30,580	**	12,500.00	2.46
2013	1,648		20,400.00	0.08
<b>Constant</b>				
	-11,230		30,700.00	-0.37

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$

N = 1,966, Adjusted R-squared = 0.11

Abutting = 243 Non-abutting = 1,723

Location on Class II bike infrastructure was not associated with a significant change in sales after installation of bike infrastructure. There was also no significant relationship between the change in on-street parking and the change in sales over time. This suggests that removing parking may not be as damaging as business owners claim. The curb management variables representing the availability of a commercial or passenger loading zone were also not significant.

The model identified a number of other predictors that did have a significant effect on the change in sales. The only corridor variables to have a significant association were dummy variables indicating the business was located on a primary or secondary road. These roads, which tend to provide state and regional connectivity, were both associated with a negative change in sales over time when compared to the reference category of neighborhood roads. In this model, neighborhood roads are low-volume roads providing local connectivity. This could indicate that larger automobile traffic counts do not necessarily translate to larger numbers of customers or higher sales. Instead, it appears that calmer local streets see higher sales.

Characteristics of the individual business proved to be some of the strongest predictors of business performance. Industry was a reliably predictive variable, though not across the board, and business age was also significant. Business age was associated with a decline in sales, indicating that older businesses had slightly lower sales on average than their newer counterparts, all else equal. Bars, service-providing businesses, and financial businesses all saw increases in sales, while automobile-oriented businesses, businesses providing health services, and businesses selling goods for the home saw declines in sales.

Neighborhood characteristics were generally insignificant, indicating that the surrounding population is a poor predictor of how a business will fare. Finally, to control for temporal fluctuations in the market, dummy variables indicating the year that served as the midpoint between the average sales before and after was also included within the model. These variables do not have a strong policy connection and serve to control for market fluctuations.

Interactions between the presence of Class II facilities and road classification were generally not significant, except for the interaction term between Class II facilities and neighborhood roads. This interaction was associated with an increase in sales by around \$20,000, indicating that businesses on low-volume neighborhood roads with Class II facilities saw their sales rise on average \$20,000 before and after the installation of bike infrastructure. Businesses on lower-volume streets may benefit more from bike infrastructure because these streets are already the most bikeable, and the addition of bike infrastructure may enable even more shoppers to access the corridor, particularly those who might have not chosen to bike otherwise.

When Class II facilities were interacted with business industry, two significant associations emerged. Automobile-oriented businesses, such as gas stations and car dealerships, and businesses selling goods for the home (e.g. furniture or carpet stores) both saw significant declines in sales. These industries did not see general significant sectoral decline, which would have been indicated with a negative coefficient on the general industry variable, which suggests that Class II infrastructure has a differing negative effect on businesses in these industries. Home goods stores tend to sell goods that may be too large to reasonably transport on a bicycle, which could be at the root of this negative association. Automobile-oriented businesses could see some customers deterred by the increased presence of cyclists on the corridor; they may see these cyclists as a complication that could be avoided by patronizing a different business on a different corridor without infrastructure.

Table 7 summarizes the model that examines the change in sales for businesses abutting Class III infrastructure against non-abutting businesses. Location on Class III infrastructure was not associated with significant change in sales over time. Location on primary and secondary roads demonstrated a negative association with change in sales of about \$19,000 and \$32,000, respectively, indicating that location on a higher auto traffic corridor was not necessarily associated with higher sales, as was seen in the model of Class II corridors.

Business characteristic variables generally showed similar patterns to those seen in Table 6, with business age and chains exhibiting negative associations with change in sales. Services, financial services, bars, and restaurants saw increases in sales, perhaps hinting at a market shift away from traditional retail outlets for purchasing goods and towards businesses offering services. Of the neighborhood variables, the only significant association was a positive change in sales for businesses in denser parts of San Francisco (more than 15 people per acre).

The interactions between bike infrastructure and road classification were not significant, and those between Class III facilities and business type were also not significant. The lack of significant associations between interactions could be due to the relatively minimal change to dynamics of the street presented by Class III infrastructure projects. These facilities may not have a very strong effect, one way or the other, on business dynamics.

Table 7: Change in sales for Class III facilities in San Francisco

	Coefficient	Sig.	Standard Error	T Statistic
<b>Roadway characteristics</b>				
Class III	52,060		53,000.00	0.98
Primary road	-18,500	**	9,179.56	-2.02
Secondary road	-32,460	***	9,982.64	-3.25
Tertiary road	-4,901		7,549.93	-0.65
<b>Business characteristics</b>				
Years Old	-1,896	***	176.95	-10.72
Chain	-32,870	***	11,400.00	-2.89
Bar	57,930	***	16,600.00	3.48
Restaurant	16,250		10,000.00	1.62
Grocery	560		13,000.00	0.04
Personal goods	2,031		10,300.00	0.20
Home goods	-10,260		14,000.00	-0.73
Services	39,790	***	9,636.91	4.13
Entertainment	42,630		64,600.00	0.66
Financial services	40,200	**	18,300.00	2.20
Health services	-13,020		9,406.37	-1.38
Automobile-oriented	-26,030		16,100.00	-1.61
<b>Bike infrastructure and road class interactions</b>				
Primary roads x Class II	-17,580		57,500.00	-0.31
Secondary roads x Class II	-9,272		54,300.00	-0.17
Tertiary roads x Class II	-63,840		52,200.00	-1.22
Neighborhood roads x Class II	-40,480		50,900.00	-0.80
<b>Bike infrastructure and business interactions</b>				
Bike x bar	-12,190		37,100.00	-0.33
Bike x restaurant	-5,681		24,600.00	-0.23
Bike x grocery	-21,840		29,000.00	-0.75
Bike x personal goods	-3,235		27,400.00	-0.12
Bike x home goods	28,120		31,200.00	0.90
Bike x services	-7,993		24,300.00	-0.33
Bike x entertainment	17,500		25,100.00	0.70
Bike x financial services	-18,020		37,200.00	-0.49
Bike x health services	-103,200		131,000.00	-0.79

Bike x automobile-oriented	40,990		40,200.00	1.02
<b>Neighborhood characteristics</b>				
Establishment density (businesses/ per acre)	-251		399.27	-0.63
Urban	17,360	*	9,025.08	1.92
Suburban	0		0.00	0.21
Median household income (\$1,000)	-54		143.32	-0.37
Percent housing renter occupied	156		205.60	0.76
Percent Latino	170		246.48	0.69
Percent Black	-213		422.68	-0.50
<b>Installation year</b>				
1996	-7,849		21,600.00	-0.36
1997	13,190		14,800.00	0.89
1998	-4,168		45,900.00	-0.09
1999	14,850		27,000.00	0.55
2000	-19,140		23,400.00	-0.82
2002	11,120		25,500.00	0.44
2003	-18,820		16,300.00	-1.16
2004	22,620		21,300.00	1.06
2005	-13,030		8,396.35	-1.55
2006	-8,434		9,746.94	-0.87
2008	7,199		36,500.00	0.20
2009	15,500		11,700.00	1.32
2011	17,290	*	9,904.18	1.75
2012	30,790	***	10,700.00	2.89
2013	1,823		20,000.00	0.09
Constant	-35,640		27,800.00	-1.28

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$

N = 1,985, Adjusted R-squared = 0.1

Abutting = 172 Non-abutting = 1,812

Overall, the sales change models for San Francisco suggest that location on bicycle infrastructure has a neutral effect on change in sales overall, with a few cases where bike infrastructure may benefit businesses and a few cases where it may have a detrimental effect. Overall, location on Class II infrastructure did not result in a significant change in sales one way or the other. However, when looking only at businesses on neighborhood roads with bike infrastructure versus all other businesses, these businesses saw an increase in sales of around \$20,000 per year. Automobile-oriented businesses and businesses selling home goods that were

located on corridors with Class II facilities did see a decline in sales of around \$100,000 per year each. Location on Class III facilities was not associated with a statistically significant change in sales. This generally disproves business owners' claims that bike infrastructure is bad for business, though it generally does not confirm cyclist advocates' claims that bike infrastructure is good for business. Instead, it appears that for businesses in San Francisco, there are a multitude of other factors that do have a determining effect on the change in sales a business sees over time.

### Alameda County

The following section presents models for businesses in Alameda County, broken down by infrastructure type. The models include the same set of independent variables that were used in the estimation above, only without the parking change variable for the Class II model, as parking data were not collected for Alameda County.

The model for Class II facilities in Alameda County is in Table 8. This model shows no significant association between location on a Class II corridor and change in sales. The model exhibited similar relationships to those discussed above in terms of business and neighborhood characteristic effects. Roadway classification was significant at the 90% confidence level for primary roads, with a decline in sales of around \$7,000.

The model shows similar relationships between business characteristics and business performance, with business age and chain status associated with negative change in sales. Automobile-oriented businesses and stores selling home goods saw declines in sales, as did grocery stores -unlike in the San Francisco models. Bars, restaurants, service-providing businesses, and financial services saw increases in sales. Dummy variables indicating an urban or suburban environment were associated with increases in sales, indicating that increased population density was associated with higher revenues.

The interaction between Class II facilities and tertiary roads was significant, with an associated increase in sales of \$20,000. Interestingly, in San Francisco, businesses on the lowest-volume streets saw the most significant increase. In Alameda County, an increase of similar magnitude was seen for businesses on slightly higher-volume tertiary roads. Across both models, bike infrastructure seemed to have a positive effect when located on lower-volume roads. The interaction between Class II facilities and automobile-oriented businesses also showed a significant negative association. Grocery stores located on Class II facilities also saw a significant decline in sales. In a counterintuitive finding, restaurants on Class II facilities saw a significant decline in sales. However, it is important to note that this model only includes businesses that were active before and after the installation of bike infrastructure. New restaurants that opened after the installation of bike infrastructure could be performing well but this model does not measure their performance.

**Table 8: Change in sales for Class II facilities in Alameda County**

	Coefficient	Sig.	Standard Error	T Statistic
<b>Roadway characteristics</b>				
Class II	9,610		8,134.17	1.18
Primary road	-7,026	*	4,113.33	-1.71
Secondary road	-1,170		3,978.49	-0.29
Tertiary road	2,837		4,269.66	0.66
<b>Business characteristics</b>				
Years Old	-1,780	***	96.33	-18.48
Chain	-59,080	***	5,782.77	-10.22
Bar	32,890	***	12,300.0,0	2.68
Restaurant	11,160	**	5,365.50	2.08
Grocery	-45,790	***	6,488.51	-7.06
Personal goods	6,038		5,526.21	1.09
Home goods	-11,490	*	6,318.97	-1.82
Services	32,030	***	4,775.27	6.71
Entertainment	41,020		41,100.00	1.00
Financial services	34,770	***	7,419.03	4.69
Health services	-7,829		4,815.75	-1.63
Automobile-oriented	-13,860	**	5,455.79	-2.54
<b>Bike infrastructure and road class interactions</b>				
Primary roads x Class II	-7,925		8,024.61	-0.99
Secondary roads x Class II	5,728		6,733.27	0.85
Tertiary roads x Class II	20,800	***	7,943.36	2.62
Neighborhood roads x Class II	-8,997		5,467.44	-1.65
<b>Bike infrastructure and business interactions</b>				
Bike x bar	18,760		38,200.00	0.49
Bike x restaurant	-30,490	**	13,500.00	-2.25
Bike x grocery	-41,060	**	17,400.00	-2.36
Bike x personal goods	4,033		15,800.00	0.26
Bike x home goods	-6,350		17,900.00	-0.35
Bike x services	-10,780		13,000.00	-0.83
Bike x entertainment	8,461		92,600.00	0.09
Bike x financial services	-17,270		21,200.00	-0.82
Bike x health services	-18,550		12,400.00	-1.50

Bike x automobile-oriented	-34,730	**	15,600.00	-2.22
<b>Neighborhood characteristics</b>				
Establishment density (businesses/ per acre)	-45		190.74	-0.24
Urban	8,431	*	4,348.76	1.94
Suburban	7,990	*	4,339.54	1.84
Median household income (\$1,000)	-34		84.01	-0.41
Percent housing renter occupied	-41		99.85	-0.41
Percent Latino	140		104.95	1.33
Percent Black	28		101.14	0.28
<b>Installation year</b>				
1997	35,660		22,600.00	1.58
1998	-5,739		20,200.00	-0.28
1999	-9,861		10,500.00	-0.94
2000	594		14,700.00	0.04
2001	-19,810	**	9,537.81	-2.08
2003	-45,710	***	9,165.18	-4.99
2004	-23,570	**	10,900.00	-2.17
2005	-35,330	***	10,300.00	-3.43
2006	-16,530		11,700.00	-1.41
2007	-30,280	***	9,356.32	-3.24
2008	-25,780	***	9,784.41	-2.64
2009	-7,841		11,800.00	-0.66
2010	-7,994		12,000.00	-0.67
2011	-8,569		12,100.00	-0.71
2012	-12,940		9,492.36	-1.36
Constant	581		15,600.00	0.04

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$

N = 8,815, Adjusted R-squared = 0.09

Abutting = 1,310 Non-abutting = 7,505

Table 9 summarizes the model estimating change in sales for businesses abutting Class III bikeways and all non-abutting businesses. The model found a negative association between Class III bicycle infrastructure and change in sales in Alameda County. However, the coefficients associated with each of the four interactions between road classification and Class III facilities are all significant, positive, and generally larger than the negative association between Class III infrastructure and change in sales. All businesses are located on one of the four corridor types. As such, the effect of Class III infrastructure on sales could be interpreted as the sum of the

Class III coefficient plus the coefficient for the interaction between bike infrastructure and the roadway classification. So businesses on primary roads with Class III facilities exhibited an increase in sales of \$23,600. Businesses on tertiary and neighborhood roads with Class III facilities also saw an increase in sales of \$16,400 and \$5,700, respectively. Businesses on secondary roads with Class III facilities saw a decline in sales of nearly \$15,000. While the coefficient for Class III facilities on its own looks large, it is generally not effectual when the interaction terms are taken into account.

As was seen in the other sales change models, business age and chains were associated with significant declines in sales. Bars, services, and financial services were associated with significant increases over time. Grocery stores, stores selling home goods, health service providers, and automobile-oriented businesses saw significant declines. Neighborhood variables were not significant, save for a dummy variable indicating a higher population density.

Interactions between industry and Class III bike infrastructure were generally insignificant. Financial service businesses on Class III facilities were associated with a decline in sales, despite a general positive trend for businesses in the financial services industry. Considering the general trends and the interaction between bike infrastructure and industry, financial service businesses on Class III facilities saw a decline in sales of about \$30,000.

**Table 9: Change in sales for Class III facilities in Alameda County**

	Coefficient	Sig.	Standard Error	T Statistic
<b>Roadway characteristics</b>				
Class III	-561,900	***	115,000.00	-4.88
Primary road	-6,834	*	3,751.69	-1.82
Secondary road	1,088		3,532.44	0.31
Tertiary road	6,582	*	3,848.75	1.71
<b>Business characteristics</b>				
Years Old	-1,816	***	92.84	-19.56
Chain	-61,630	***	5,506.37	-11.19
Bar	34,030	***	11,600.00	2.94
Restaurant	5,672		4,868.68	1.17
Grocery	-51,960	***	5,982.52	-8.69
Personal goods	6,347		5,148.87	1.23
Home goods	-12,190	**	5,880.44	-2.07
Services	30,590	***	4,410.72	6.94
Entertainment	41,220		36,600.00	1.13

Financial services	33,000	***	6,905.53	4.78
Health services	-10,370	**	4,395.80	-2.36
Automobile-oriented	-17,660	***	5,076.47	-3.48
<b>Bike infrastructure and road class interactions</b>				
Primary roads x Class III	585,500	***	117,000.00	5.02
Secondary roads x Class III	547,000	***	117,000.00	4.69
Tertiary roads x Class III	578,300	***	116,000.00	4.97
Neighborhood roads x Class III	567,600	***	116,000.00	4.89
<b>Bike infrastructure and business interactions</b>				
Bike x bar	-29,280		50,200.00	-0.58
Bike x restaurant	-23,930		19,800.00	-1.21
Bike x grocery	40,960		26,900.00	1.52
Bike x personal goods	-31,160		22,300.00	-1.40
Bike x home goods	-3,498		27,000.00	-0.13
Bike x services	-15,080		19,900.00	-0.76
Bike x entertainment	-8,627		16,700.00	-0.52
Bike x financial services	-64,690	**	26,500.00	-2.44
Bike x automobile-oriented	26,540		23,000.00	1.16
<b>Neighborhood characteristics</b>				
Establishment density (businesses/ per acre)	24		180.65	0.13
Urban	7,194	*	4,155.62	1.73
Suburban	5,762		4,156.65	1.39
Median household income (\$1,000)	-39		81.84	-0.48
Percent housing renter occupied	-71		96.09	-0.74
Percent Latino	134		94.85	1.42
Percent Black	93		93.71	1.00
<b>Installation year</b>				
2001	-17,870	***	5,437.38	-3.29
2003	-42,030	***	4,808.95	-8.74
2004	-20,250	***	7,652.33	-2.65
2005	-31,420	***	6,502.60	-4.83
2006	-15,290	*	8,783.90	-1.74
2007	-26,980	***	5,074.68	-5.32
2008	-23,300	***	5,734.33	-4.06
2009	-8,979		8,551.36	-1.05
2010	-6,913		9,000.29	-0.77
2011	-4,305		9,006.22	-0.48

2012	-9,069	*	5,085.82	-1.78
Constant	573		13,700.00	0.04

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$

N = 9,377, Adjusted R-squared = 0.1

Abutting = 562 Non-abutting = 8,815

To summarize, in Alameda County, bike infrastructure did not generally have significant effects, with a few exceptions. The models suggest that Class II infrastructure has no significant effect on sales. Class III infrastructure was associated with a decline in sales, though this effect generally flipped to an increase in sales for businesses on primary, tertiary, and neighborhood roads. Class III infrastructure on secondary roads did have an overall negative effect on sales.

In Alameda County, as in San Francisco, business characteristics were the overall most reliable predictors of sales. Neighborhood characteristics were poor predictors. Corridor characteristics were somewhat predictive, with primary roads associated with a decline in sales in Class II and III models for Alameda County.

Neither corridor nor neighborhood characteristics were reliably predictive across models. Population density was occasionally significant, with higher population densities associated with positive changes in sales.

The models suggest that business characteristics – something planners have little to no control over - overwhelmingly predict business performance. While bicycle infrastructure generally does not have a blanket positive effect on sales, as cycling advocates may like to say, it also does not have an overall negative effect on businesses. Instead, there are other factors that have a much larger influence on how a business fares.

### Turnover models

Business turnover is another good metric for business performance. Understanding whether bicycle infrastructure affects the likelihood that a business will close is relevant to business owners and planners alike.

In this work, abutting businesses were coded as having turned over if their last year in the NETS dataset was within three years of the date of installation for the bicycle infrastructure project. Businesses that closed more than three years after the infrastructure had been installed, or that were marked as open in 2014, were coded as having remained open. Non-abutting businesses were joined to the nearest corridor with bicycle infrastructure, and turnover was calculated in a similar way to compare abutting to non-abutting businesses.

Abutting businesses that closed before the infrastructure was installed were omitted from analysis. Moves out from the corridor are another form of business

turnover, though they represent a small fraction of all businesses in the NETS dataset. Businesses that moved were excluded from analysis, though further research could incorporate a multinomial logistic regression to estimate the decision to move, close, or remain open.

A binary logistic regression on the outcome of 1 (closing) and 0 (remaining open) was conducted to identify statistically significant predictors of business closure. The results are presented in odds ratio format for ease of interpretation. Odds ratio coefficients that are greater than one indicate an increase in the likelihood that the business closes. Odds ratio coefficients that are less than one indicate a decrease in the likelihood that a business closes. Like the models of sales change, models were estimated for each infrastructure type within each county to account for heterogeneous relationships that may exist between different types of bicycle infrastructure.

### San Francisco County

As with the sales change models, the turnover models include characteristics of the business, the neighborhood, and the corridor, including the change in number of on-street parking spaces for Class II facilities. The models also include dummy variables for the installation year to control for temporal patterns.

Abutting a Class II facility did not have a significant relationship with the likelihood of a business in San Francisco turning over. Similarly, the presence of a passenger loading zone or the change in the number of on-street parking spaces also did not exhibit a statistically significant impact. Table 10 summarizes the full model output. Businesses on high-volume primary roads were more likely to turn over, though this was the only corridor-specific variable that demonstrated a statistically significant relationship. Older businesses and larger businesses were less likely to turn over, as might be expected. Similarly, grocery stores and businesses providing health services were significantly less likely to turn over. Establishment density was associated with a reduction in the likelihood of turning over, though no other neighborhood variables were statistically significant.

**Table 10: Likelihood of turnover for Class II facilities in San Francisco**

	Odds Ratio	Sig.	Z Statistic
<b>Roadway characteristics</b>			
Class II	0.77		-1.16
Primary road	1.42	*	1.93
Secondary road	1.00		0.01
Tertiary road	0.90		-0.65
Change in on street parking	1.43		1.34
Passenger loading zone	0.83		-0.31

Commercial loading zone	0.00		-0.01
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### Business characteristics

Number of employees (last year)	0.87	***	-3.69
Sales (\$1,000, last year)	1.00	***	5.55
Years Old	0.98	***	-4.14
Chain	0.84		-0.58
Bar	1.00		0.00
Restaurant	0.80		-0.87
Grocery	0.29	***	-3.83
Personal goods	0.79		-1.21
Home goods	0.93		-0.29
Services	0.94		-0.27
Entertainment	0.00		0.00
Financial services	0.83		-0.48
Health services	0.41	***	-4.87
Automobile-oriented	0.65		-1.36

### Neighborhood characteristics

Establishment density (businesses/ per acre)	0.97	***	-3.41
Urban	1.05		0.25
Median household income (\$1,000)	1.00	***	-2.58
Percent Latino	1.00		-0.49
Percent Black	0.99		-1.13

### Installation Year

1996	0.19	***	-3.75
1997	0.33	***	-4.17
1998	0.22	**	-2.43
1999	0.12	***	-3.52
2000	0.34	***	-2.78
2001	0		0.00
2002	0.16	***	-3.83
2003	0.17	***	-4.65
2004	0.28	**	-1.99

2005	0.25	***	-6.46
2006	0.52	***	-2.75
2008	0.86		-0.18
2009	0.72		-1.18
2011	0.68	*	-1.74
2012	0.22	***	-3.82
2013	0		0.00

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$   
N = 9,377, Adjusted R-squared = 0.09

Abutting a Class III facility also did not register a statistically significant association with an increased likelihood of closure. Table 11 shows the results of the model estimation for Class III facilities in San Francisco. As seen in the previous model, location on primary roads was associated with an increase in the likelihood of a business turning over. The other trends related to business characteristics also held true, with older and larger businesses seeing a reduced likelihood of turning over. This model differs from the model of Class II facilities in that it registers a strong association between the share of black residents and the likelihood that a business turns over.

**Table 11: Likelihood of turnover for Class III facilities in San Francisco**

	Odds Ratio	Sig.	Z Statistic
<b>Roadway characteristics</b>			
Class III	0.89		-0.81
Primary road	1.33	**	1.79
Secondary road	0.97		-0.17
Tertiary road	0.8		-1.64
<b>Business characteristics</b>			
Number of employees (last year)	0.87	***	-4.09
Sales (\$1,000, last year)	1.00	***	4.68
Years Old	0.98	***	-4.13
Chain	1.00		-0.02
Bar	1.09		0.28
Restaurant	1.01		0.05
Grocery	0.51	***	-2.68
Personal goods	1.29		1.42
Home goods	1.37		1.50

Services	1.03		0.11
Entertainment	2.61		0.83
Financial services	0.80		-0.62
Health services	0.63	***	-2.76
Automobile-oriented	1.14		0.52
<b>Neighborhood characteristics</b>			
Establishment density (businesses/ per acre)	0.99	**	-1.79
Urban	1.06		0.31
Suburban	1.23		0.37
Median household income (\$1,000)	1		0.47
Percent Latino	0.72		-0.54
Percent Black	8.82	***	2.65
<b>Installation Year</b>			
1996	0.10	***	-4.17
1997	0.13	***	-5.19
1998	0.09	***	-3.49
1999	0.06	***	-4.51
2000	0.06	***	-5.12
2001	0.00		0.00
2002	0.04	***	-4.91
2003	0.04	***	-5.41
2004	0.06	***	-4.88
2005	0.10	***	-6.52
2006	0.28	***	-3.47
2008	0.35		-1.23
2009	0.23	***	-3.40
2010	0.37	***	-2.99
2011	0.27	***	-3.39
2012	0.10	***	-5.20
2013	0.00		0.00

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$   
N = 3,806, Pseudo R-squared = 0.09

The models for both Class II and Class III infrastructure in San Francisco indicate no significant association between location on bicycle infrastructure and the likelihood of a business turning over. While other characteristics of the business, the

neighborhood, and the corridor seem to have an effect, it seems that bicycle infrastructure may not play such a deterministic role in whether or not a business closes.

### Alameda County

The following set of models describes the relationship between abutting bicycle infrastructure and the likelihood of a business closing in Alameda County. The independent variables are the same as in the San Francisco turnover models, save for the exclusion of on-street parking variables.

Abutting a Class II facility decreases the likelihood of a business closing at a weakly significant (90% confidence) level (Table 12). Businesses are about 22% less likely to close if they are located on Class II infrastructure, all else equal. Unlike in San Francisco, location on a primary road was associated with a decrease in the likelihood of a business closing, while other road types showed no significant association. This inversion in trends between primary roads in San Francisco and Alameda County could be influenced by the relative auto-orientedness of Alameda County, which encompasses a much wider array of densities than San Francisco. Direct access to high volume automobile corridors may actually be a positive for businesses in Alameda County, while businesses in San Francisco may have suffered due to the availability and attractiveness of other modes.

The model for Class II infrastructure otherwise presented a similar picture to what was seen in previous models in this section. Business age and number of employees were associated with a decrease in the likelihood of a business closing. Grocery stores, personal goods stores, and providers of health services were less likely to close, while service industry businesses were more likely to close.

**Table 12: Likelihood of turnover for Class II facilities in Alameda County**

	Odds Ratio	Sig.	Z Statistic
<b>Roadway characteristics</b>			
Class II	0.78	*	-1.80
Primary road	0.76	***	-3.31
Secondary road	0.94		-0.85
Tertiary road	0.92		-1.04
<b>Business characteristics</b>			
Number of employees (last year)	0.87	***	-9.09
Sales (\$1,000, last year)	1.00	***	14.79
Years Old	0.93	***	-20.28
Chain	1.17		1.25

Bar	1.24		0.97
Restaurant	0.97		-0.24
Grocery	0.69	***	-3.13
Personal goods	0.89	*	-1.15
Home goods	0.98		-0.23
Services	1.27	**	2.30
Entertainment	1.23		0.25
Financial services	0.95		-0.34
Health services	0.55	***	-6.69
Automobile-oriented	0.91		-0.91

### Neighborhood characteristics

Establishment density (businesses/ per acre)	1.00		0.26
Urban	1.10		1.16
Suburban	1.06		0.67
Median household income (\$1,000)	1.00		-0.07
Percent Latino	1.04		0.17
Percent Black	1.61	*	1.87

### Installation year

1997	0.21	***	-4.00
1998	0.21	***	-5.42
1999	0.22	***	-8.14
2000	0.16	***	-6.10
2001	0.2	***	-8.99
2003	0.21	***	-8.99
2004	0.22	***	-8.02
2005	0.27	***	-6.75
2006	0.61	***	-2.65
2007	0.59	***	-3.26
2008	0.82		-1.22
2009	0.53	***	-3.12
2010	0.78		-1.23
2011	0.46	***	-3.26
2012	0.28	***	-6.53
2013	0		-0.01

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$

N = 14,504, Pseudo R-squared = 0.12

The final model, presented in Table 13, shows the factors influencing the likelihood of a business closing for businesses abutting Class III infrastructure in Alameda County. Location on a corridor with Class III infrastructure is associated with a decline in the likelihood of turnover (albeit at a 90% confidence level). Businesses on primary roads are similarly less likely to turn over, as was seen in the Class II model. Business characteristic patterns exhibited in the Class II model were replicated in the Class III model. Additionally, the share of black residents in the surrounding neighborhood was associated with a slight increase in the likelihood of a business closing.

**Table 13: Likelihood of turnover for Class III facilities in Alameda County**

	Odds Ratio	Sig.	Z Statistic
<b>Roadway characteristics</b>			
Class III	0.78	*	-1.80
Primary road	0.76	***	-3.31
Secondary road	0.94		-0.85
Tertiary road	0.92		-1.04
<b>Business characteristics</b>			
Number of employees (last year)	0.87	***	-9.09
Sales (\$1,000, last year)	1.00	***	14.79
Years Old	0.93	***	-20.28
Chain	1.17		1.25
Bar	1.24		0.97
Restaurant	0.97		-0.24
Grocery	0.69	***	-3.13
Personal goods	0.89		-1.15
Home goods	0.98		-0.23
Services	1.27	**	2.30
Entertainment	1.23		0.25
Financial services	0.95		-0.34
Health services	0.55	***	-6.69
Automobile-oriented	0.91		-0.91
<b>Neighborhood characteristics</b>			
Establishment density (businesses/ per acre)	1.00		0.26

Urban	1.10		1.16
Suburban	1.06		0.67
Median household income (\$1,000)	1.00		-0.07
Percent Latino	1.04		0.17
Percent Black	1.61	*	1.87
<b>Installation year</b>			
1997	0.21	***	-4.00
1998	0.21	***	-5.42
1999	0.22	***	-8.14
2000	0.16	***	-6.10
2001	0.2	***	-8.99
2003	0.21	***	-8.99
2004	0.22	***	-8.02
2005	0.27	***	-6.75
2006	0.61	***	-2.65
2007	0.59	***	-3.26
2008	0.82		-1.22
2009	0.53	***	-3.12
2010	0.78		-1.23
2011	0.46	***	-3.26
2012	0.28	***	-6.53
2013	0		-0.01

Sig. \* =  $p < 0.10$ , \*\* =  $p < 0.05$ , \*\*\* =  $p < 0.00$

N = 13,504, Pseudo R-squared = 0.12

The turnover models suggest that location on bike infrastructure has either no significant effect or a pro-business effect. This differs from the picture that emerged from the sales over time regressions. This being said, characteristics related to the business itself were still reliably predictive factors, which means that infrastructure is not the only determinant of outcomes for businesses.

### Summary

The models presented above seek to control for characteristics of the corridor, neighborhood, and business itself to illuminate the effect of bicycle infrastructure in isolation. All else equal, the relationship between location on bicycle infrastructure and business success is mixed, with bike infrastructure appearing to have a neutral effect on change in sales and a neutral or positive effect on likelihood of turnover.

In San Francisco and Alameda Counties, location on Class II facilities was not associated with a significant change in sales, except for on neighborhood roads, which saw an increase. Location on Class III facilities in San Francisco was not

associated with a significant change in sales. In Alameda County, location on Class III infrastructure was associated with a generally positive change in sales, though in the case of facilities on secondary roads, there was a negative association.

Suggested table:

	<b>Alameda County</b>	<b>San Francisco County</b>
Change in Sales		
Class I	+	-
Class II	+	
Class III	-	
Turnover		
	Etc.	Etc.

The control variables also provided unique insights. Business characteristics, particularly industry, were reliably significant predictors. Across both counties, bars, services, and financial services all registered significant increases in sales. Meanwhile, automobile-oriented businesses saw significant declines in sales. Older businesses generally saw decreases in sales, as did chains.

Neighborhood characteristics were generally not significant. Businesses in higher density urban environments tended to see increases in sales. In San Francisco, location on primary and secondary roads was associated with a decline in sales when compared against the reference category, neighborhood roads. Location on primary roads was found to have a weak negative association for businesses on Class II infrastructure in Alameda County, though no effect was found for the Class III model in Alameda County. This suggests that high automobile volumes do not necessarily translate to increased sales, especially in San Francisco. In San Francisco, the model for Class II facilities showed no significant association between parking removal and sales, indicating that parking removal may not be detrimental to business performance.

In terms of turnover, both classes of bike infrastructure in San Francisco showed no significant relationship with the likelihood of turnover. Change in on-street parking was also not significant in the Class II model for San Francisco. In Alameda County, businesses abutting Class II and Class III corridors saw a lower likelihood of turnover. The models suggest that bicycle infrastructure either has no effect or has a positive effect on business outcomes by not affecting or reducing the likelihood of turnover.

In San Francisco, the business age and number of employees were associated with a decline in the likelihood of turnover, as were certain industries, including grocery stores and health care service providers. Additionally, establishment density tended to be associated with a reduced likelihood of turnover. In Alameda County, similar trends held true, with the likelihood of a business closing also decreasing as businesses grew in size and age. Additionally, location on a primary road reduced the likelihood of closure for businesses in Alameda County. Businesses that

provided services were more likely to close in Alameda County, while grocery stores and healthcare providers were less likely. As was seen in sales change models, neighborhood characteristics were rarely predictive of the likelihood of a business turning over.

Interesting contrasts emerge when comparing the results of the sales change models and the turnover models. For example, while service industry businesses generally saw an increase in sales over time, they also saw an increase in the likelihood of turnover. Conversely, grocery stores saw significant decreases in sales, but a strong decrease in the likelihood of closing. The variable representing sales in the business' last year of operation was significant in all of the turnover models, though the magnitude of the effect was small, even when converted into \$1,000 increments.

In sum, the models suggest that bicycle infrastructure has a generally mixed effect on the change in sales over time on an individual business level, with the most positive effects occurring on neighborhood-type roads. The relationship between business turnover and bicycle infrastructure is either neutral (in the case of San Francisco) or pro-business (in the case of Alameda County), with location on bicycle infrastructure of all types resulting in a reduced likelihood of turnover.

## Intercept survey of shoppers

### Introduction

This section presents findings from an intercept survey of shoppers conducted in San Francisco, Oakland, and Berkeley. The survey was conducted on two matched pair corridors, each consisting of one corridor with a Class II bike facility and one corridor with no bike infrastructure. The survey results provide some insight into how the presence of bicycle infrastructure affects consumers' mode choice decisions for travel to a commercial corridor and the differences in purchasing behavior of users of different modes, on corridors with and without bicycle infrastructure.

### Methodology

#### Site selection process

The survey was conducted at four locations, forming two matched pair sets. The first parameter of survey site selection was having one site in San Francisco and the second in Alameda County to complement analysis of secondary data and to provide a deeper picture of the relationship between bicycle infrastructure and consumer behavior in the Bay Area. With this parameter in place, we then examined a number of potential survey corridors in San Francisco and Alameda Counties, evaluating potential sites based primarily on storefront establishment density. Corridors with bicycle infrastructure installed after 2014 were excluded so intercept survey data could be contextualized by data from NETS. This excluded high-density commercial corridors in the East Bay such as Telegraph Avenue in downtown Oakland or Piedmont Avenue in North Oakland. Ultimately, we selected Valencia Street in San Francisco's Mission District and Broadway in north Oakland as the study corridors with bicycle infrastructure.

Researchers then used suitability-type analyses in ArcGIS to identify matches for the two pre-defined corridors with bicycle lanes. The match identification process was limited to the county within which the matching corridor with bike infrastructure fell. Corridors were characterized using demographic data about the residents in the surrounding area from the 2014 American Community Survey and data about establishment density and industry mix from NETS.

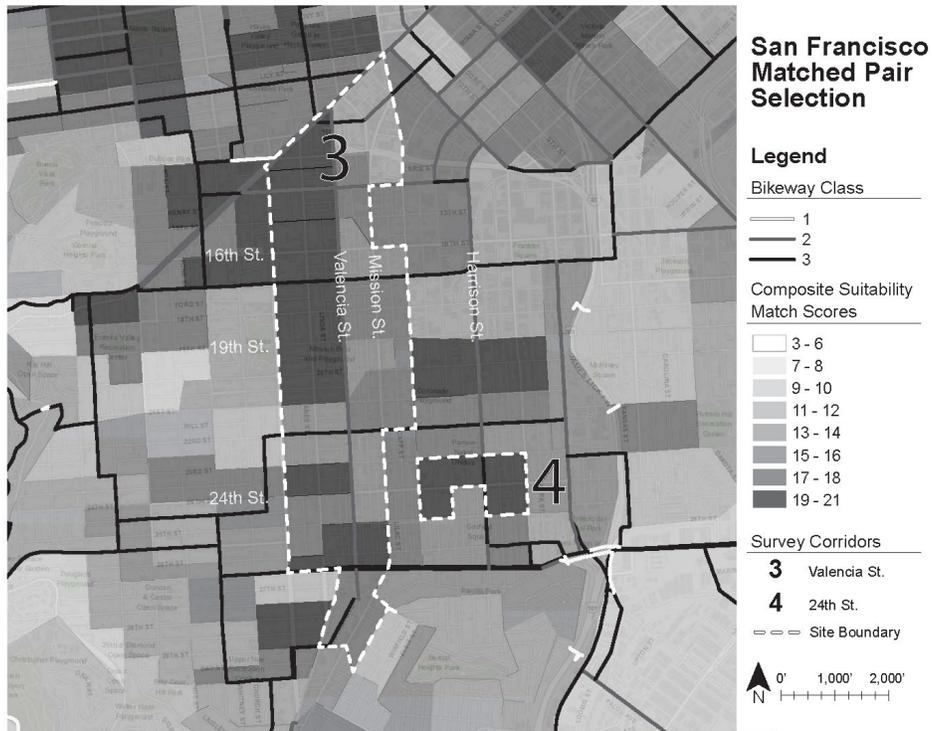
The corridor average and standard deviation was calculated for each criterion at the Census Block Group level. Census Block Groups in the study county were then weighted based on their distance from the corridor mean. Block Groups received a weight of three for a criterion if their mean was within one standard deviation of the corridor mean. Block Groups received two points for a criterion if they were between one and two standard deviations from the corridor mean, and one point if they were between two and three standard deviations. The weights for each criterion were summed together to produce a composite weight. We used the composite weights in conjunction with qualitative assessments of the various potential comparison corridors to select a fitting match.

Table 14 shows the criteria that were included in the matched pair identification process and the average and standard deviation for the corridors with bicycle infrastructure. Figures 7 and 8 show the suitability analysis for Valencia Street and Broadway, respectively.

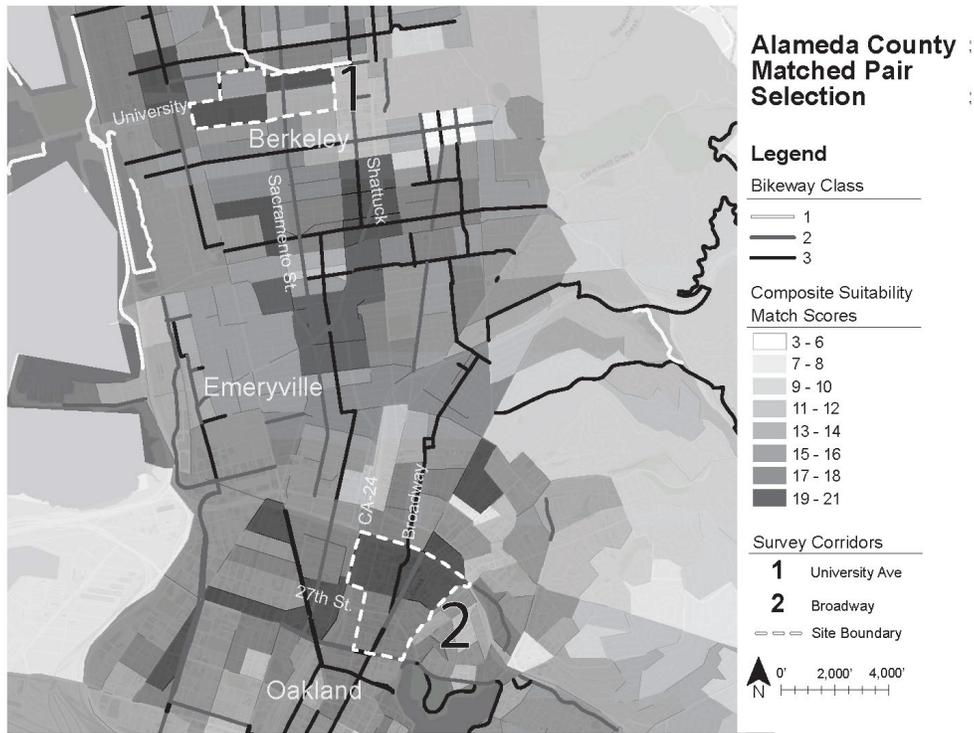
**Table 14: Matched pair selection criteria**

	Valencia Street		Broadway	
	Mean	S.D.	Mean	S.D.
<b>Commercial characteristics</b>				
Establishment density (est./acre)	6.1	2.1	3.3	1.1
Percent storefront	19.5%	.084	8.9%	.053
Percent eating/drinking establishments	8.3%	.049	3.7%	.025
<b>Neighborhood characteristics</b>				
Population density (ppl./acre)	61.6	18.0	22.1	8.4
Median household income	80,693	34,887	39,780	12,722
Auto commute mode share	25.1%	7.3	42.5%	10.2
Bike commute mode share	8.2%	4.6	10.5%	5.2

**Figure 7 Suitability analysis for Valencia Street**



**Figure 8: Suitability analysis for Broadway**



**Survey sites**

Surveys were conducted at the four Bay Area locations described above. Figure 9 shows a map of the survey sites. The following section briefly describes commercial conditions and transportation amenities on the corridors.

Figure 9: Survey sites



### Broadway

Broadway is located in northeast Oakland and serves as the dividing line between Pill Hill, to the West, and Westlake, to the east. The corridor features two travel lanes in each direction, separated by a grassy median. In 1998, Class II facilities were installed on both directions. The corridor also has metered on-street parking and is served by AC Transit's 51A bus route.

The corridor has a moderate commercial density when compared to Oakland as a whole, with some auto-oriented business uses, as well as new commercial construction near the Kaiser Permanente Hospital, including a Sprouts grocery store. The Broadway Avenue Class II facility becomes a Class III facility south of 25<sup>th</sup> Street, as it nears downtown.

Figure 10: Broadway



Image source:

<http://www2.oaklandnet.com/government/o/PWA/o/EC/s/BicycleandPedestrianProgram/OAK043750>

### **University Avenue**

University Avenue connects UC Berkeley, in the east, to predominantly residential neighborhoods in the west. University Avenue has twomedian-separated travel lanes in each direction with metered on-street parking. The corridor is served by AC Transit's 51B and 52, as well as weekday commute period service to San Francisco via the AC Transit TransBay FS route. There is no bicycle infrastructure on the corridor.

The corridor consists of medium-density mixed residential and commercial development and some auto-oriented uses at the western extent. Residential density decreases as distance from UC Berkeley increases. Two blocks to the north of University is the Hearst Avenue Class II bikeway, which provides a connection to the Class I Ohlone GreenwayThe Virginia Avenue bike boulevard is five blocks north of University and the Channing Way bike boulevard is fourblocks south.

Figure 11: University Avenue



Image source: <http://www.berkeleyside.com/2015/01/23/cyclist-hurt-5-elevator-rescues-in-berkeley-power-outage>

### ***Valencia Street***

Valencia Street is a mixed-use commercial and residential corridor in San Francisco's Mission District. Considered one of the city's primary bicycle arteries, Valencia has featured bidirectional Class II bike facilities since 1998. Bike advocates regularly impress the importance of this corridor and have advocated for more intensive bike infrastructure interventions in recent months (Rudick 2017). The corridor features three travel lanes and metered on-street parking. There is no transit service on Valencia, but the corridor is one block west of Mission Street, which has a high transit level of service and regional connection via BART.

Figure 12: Valencia Street



Image source: <http://www.missionmission.org/tag/valencia-street/>

### **24<sup>th</sup> Street**

Valencia Street's matched pair is 24<sup>th</sup> Street. 24<sup>th</sup> Street is located in the southern portion of the Mission and intersects with Valencia. The study area for 24<sup>th</sup> Street is the dense commercial portion to the east of Folsom Street, roughly a quarter-mile from Valencia Street. 24<sup>th</sup> Street has one travel lane in each direction and metered parking on either side of the street. The corridor is served by Muni bus routes 48 and 67 and does not currently include bicycle infrastructure. Development along 24<sup>th</sup> Street is medium-density mixed commercial and residential uses.

Figure 13: 24th Street



Image source: <https://www.sfgate.com/style/article/Welcome-to-Top-Shops-2015-6504145.php#next>

### Survey protocol

Surveys were conducted from November 2017 to February 2018. Surveyors were in the field during high-yield periods: evening commute hours (4pm-6pm) during weekdays, as well as weekend mornings (11am-1pm) and evenings (4pm-6pm). Surveyors were dispatched in teams of two to survey at the same site. The survey was conducted using the Qualtrics survey platform loaded onto iPads and was available in English and Spanish.

In the interest of reducing the length of the survey, questions pertained to the trip the person was currently undertaking. As such, people who were intercepted in the middle of a trip for a purpose other than shopping were excluded from the sample. Respondents that lived outside the Bay Area were also excluded.

### Questionnaire

The survey consisted of 20 questions regarding the respondent's current shopping trip, travel to the corridor, and general demographic information. Respondents answered questions on the travel mode they used to arrive at the corridor, the frequency with which they traveled to the corridor, and the frequency with which they traveled to the corridor by bike. In terms of questions about the shopping trip, respondents were asked the type of business they intended to patronize (eating and drinking place, service, entertainment, etc.) and were asked to project the amount of

money they would spend on the corridor that day. Information about the respondent, including access to a car or bicycle, income, and home location were also collected.

Overall, the survey took an average of five minutes to complete and was completed by just over 300 respondents. This translated to roughly 250 valid responses after the data cleaning process was complete and non-shoppers and non-Bay Area residents had been excluded.

## Results

Table 15 summarizes the number of responses collected at each survey location. The corridors with bike infrastructure (Valencia and Broadway) were both oversampled in an attempt to increase the number of cyclist responses and better understand cyclists' decisions. There was a relatively small number of responses for 24<sup>th</sup> Street. In total, there were 241 completed surveys, though not every respondent completed each question.

**Table 15: Responses per survey location**

	Responses
24th	19
Valencia	56
University	65
Broadway	110
<b>Total</b>	<b>250</b>

The survey seeks to tease out differences in shopper behavior at the different matched pair locations. To begin, respondents were asked what mode they used to arrive at the corridor that day. The 9% of respondents who used more than one mode, an e-hailing company, or selected a fill-in-the-blank "other" response were condensed into category 'other.' Table 16 presents the breakdown of various modes used by shoppers intercepted on the corridors. Of note is the relative prominence of users of 'other' modes on University Avenue and Valencia Street. Many of these respondents combined transit with another mode (drive or walk) to arrive at University, while there was a higher presence of e-hailing vehicle riders on Valencia.

**Table 16: Mode share at survey locations**

	Bike	Walk	Transit	Car	Other	Total
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24th	N	0	13	2	3	1	<b>19</b>
	%	0%	68%	11%	16%	5%	<b>100%</b>
Valencia	N	2	25	3	18	8	<b>56</b>
	%	4%	45%	5%	32%	14%	<b>100%</b>
University	N	12	26	4	5	9	<b>56</b>
	%	21%	46%	7%	9%	16%	<b>100%</b>
Broadway	N	22	30	1	53	4	<b>110</b>
	%	20%	27%	1%	48%	4%	<b>100%</b>
<b>Total</b>		<b>36</b>	<b>94</b>	<b>10</b>	<b>79</b>	<b>22</b>	<b>241</b>

Roughly the same share of shoppers biked to Broadway and University Avenue (20% and 21%, respectively). No respondents on 24<sup>th</sup> Street had biked that day, while two respondents were intercepted after biking to Valencia, representing 4%.. Nearly half of respondents drove to Broadway, perhaps encouraged by the availability of a large parking structure near the survey site. Walking was a popular mode for all survey sites.

Research suggests that cyclists are more frequent shoppers than people using other modes (Clifton 2012). A question remains as to whether or not bike infrastructure encourages people to bike more frequently for different trip types, including shopping. The survey asked respondents to report the frequency with which they traveled to the corridor by bike to make purchases. Table 17 summarizes the responses. Weekly biking was higher on the control streets (24<sup>th</sup> Street and University Avenue) than on Valencia and Broadway. This suggests that overall demographic or business mix may override the presence of bike infrastructure when influencing behavioral characteristics of a street's shoppers. Shoppers in San Francisco were more likely to never bike than people intercepted in the East Bay. This could be due in part to the Mission District's role as a regional destination, attracting people from beyond a reasonable biking distance to shop. The East Bay survey sites returned large shares of monthly riders.

Interestingly, a relatively high share of respondents reported never biking on the corridors with bike infrastructure. Overall, these findings could suggest preferences in different kinds of cyclists. Perhaps infrequent bike shoppers are generally less comfortable biking, and prefer the added feeling of safety and comfort that the bike lanes provide. Conversely, weekly bike shoppers who may feel more comfortable riding in mixed traffic may feel fine biking to corridors without infrastructure to shop. It is also possible the corridors with the highest volumes of bike users are also the busiest, and therefore the most difficult on which to install bike infrastructure.

**Table 17: Frequency of bike trips by corridor**

		Weekly	Monthly	Never	Total
24th	N	6	1	12	<b>19</b>
	%	31%	5%	63%	<b>99%</b>

Valencia	N	9	6	41	<b>56</b>
	%	16%	11%	73%	<b>100%</b>
University	N	13	10	32	<b>55</b>
	%	24%	58%	18%	<b>100%</b>
Broadway	N	11	26	73	<b>110</b>
	%	10%	66%	24%	<b>100%</b>
<b>Total</b>		<b>30</b>	<b>43</b>	<b>158</b>	<b>240</b>

Research has also shown that while cyclists may spend less per trip than shoppers traveling by other modes, the increased frequency of cyclists' shopping trips can make them greater absolute spenders than other road users (Clifton 2012, Bent and Singa 2009). As such, respondents were asked to estimate the amount of money they would spend during the day's shopping trip on the corridor. Table 18 summarizes the average expected expenditure by users of each mode. Cyclists on Broadway expected to spend twice that of the cyclists on University, suggesting that more affluent cyclists may patronize businesses on corridors with bike lanes. Such a comparison is not possible for 24<sup>th</sup> and Valencia, since no bike shoppers were intercepted on 24<sup>th</sup>. Relatively high spending on the streets with bike infrastructure may be influenced by a number of factors. Bike infrastructure may attract more boutique, high-end businesses. Alternatively - or in addition - planners may choose to locate bike lanes in areas with more expensive businesses.

Also of note are the trends in spending for users of other modes. On Broadway, transit users expected to spend almost \$100, an average larger than that for users of any other mode on the corridor by a factor of three. On Valencia, users of cars and 'other' modes had the highest estimated spending, while cyclists had the lowest. This pattern was reversed on 24<sup>th</sup>, with transit users having the highest expected spending, followed by pedestrians. 24<sup>th</sup> Street's easy accessibility to BART via the 24<sup>th</sup> Street station could have an effect by bringing in regional customers making a destination-type shopping trip and looking to spend more.

**Table 18: Estimated expenditure by mode and survey location**

	Bike	Walk	Transit	Car	Other	<b>Average</b>
24th	-	\$36	\$60	\$28	\$20	<b>\$26</b>
Valencia	\$20	\$35	\$53	\$59	\$67	<b>\$49</b>

University	\$15	\$32	\$15	\$23	\$78	<b>\$27</b>
Broadway	\$33	\$23	\$96	\$38	\$25	<b>\$33</b>
<b>Average</b>	<b>\$25</b>	<b>\$30</b>	<b>\$43</b>	<b>\$41</b>	<b>\$52</b>	<b>\$35</b>

To test the relationship between frequency of trips and spending, we also tabulated average expenditure at the survey sites by frequency of biking. Table 19 shows averages for each survey site broken down by how frequently the respondent comes to the corridor by bike to make purchases. Shoppers who biked to University weekly to make purchases had the highest estimated expenditure by nearly a factor of two, with never bikers having the lowest. On Broadway, monthly bike shoppers had the highest expenditure, followed by weekly bikers and then never bikers. In San Francisco, shoppers who bike to Valencia weekly to make purchases estimated that they would spend \$38, while shoppers who never biked had the highest estimated spending at \$55. On 24<sup>th</sup> Street, shoppers who bike weekly estimated that they would spend \$48, followed by monthly and then never bikers. On both corridors without bicycle infrastructure, regular bike shoppers had the highest estimated spending.

By and large, Table 19 suggests that more frequent bikers are more likely to spend more, though this pattern was not seen on 24<sup>th</sup> Street. Never bikers had the lowest average spending on three of the corridors, with the exception being Valencia, where they reported the highest estimated spending.

**Table 19: Estimated expenditure by frequency of biking**

Survey site	Estimated expenditure
Broadway Weekly	\$33
Monthly	\$46
Never	\$27

University	Weekly	\$40
	Monthly	\$24
	Never	\$22
Valencia	Weekly	\$38
	Monthly	\$30
	Never	\$55
24th	Weekly	\$48
	Monthly	\$40
	Never	\$31
<b>Average</b>		<b>\$35</b>

### Summary

The survey results explore the complex relationship between built environment and consumer behavior. By surveying on matched pair corridors with and without bike lanes, the data represent how behavior differs when consumers are presented with a commercial corridor that includes bike infrastructure. Shopper behavior is analyzed along four primary definitions: mode share, frequency of trips, estimated expenditure, and estimated expenditure by frequency of trips.

In terms of mode share, corridors with and without bike infrastructure featured similar shares of customers arriving by bike. Survey locations in the East Bay saw substantially higher shares of shoppers arriving by bike in the survey sample. However, in terms of frequency, corridors without bicycle infrastructure registered higher rates of respondents regularly shopping by bicycle. Counterintuitively, corridors with bike infrastructure saw larger shares of shoppers who never bike.

Shoppers were also asked to estimate the amount of money they would spend on the corridor that day. Findings were mixed, with more spent on corridors with bike infrastructure – but not necessarily by bikers. More frequent bikers tended to spend more.

In sum, the analysis examining differences in consumer behavior on corridors with and without bike infrastructure did find some interesting differences. However, the survey findings do not establish that bike infrastructure itself makes a difference; rather, it seems like shoppers travelling via bike may be more affluent, with greater total expenditures at local businesses.

The question of cyclist comfort was not addressed by this survey, though the differences in usage patterns on corridors with and without bike infrastructure point to comfort having an influence on infrequent riders. Future survey research

should seek to quantify comfort, though this is an admittedly personal definition that may be hard to measure objectively across respondents.

## Discussion

This research was ultimately motivated by the perceived conflict between the business community and cycling advocates. Business owners have been some of the most vocal opponents to bicycle infrastructure projects, and their opposition carries substantial weight with planners and elected officials. However, there is a relative paucity of research that provides an empirical look into the relationship between bicycle infrastructure and business performance.

Our research found that location on bicycle infrastructure is related to mixed business performance outcomes. In terms of change in sales, location on Class II facilities was not associated with a change in sales, while location on Class III facilities was neutral in San Francisco and mixed in Alameda County. Interaction terms provided additional insights, with automobile-oriented businesses located on corridors with Class II facilities seeing a significant decline in sales. Class II facilities also had significant positive effects on sales for businesses on lower-volume roads (neighborhood roads in San Francisco and tertiary roads in Alameda County). The models of the likelihood of business turnover showed that location on bicycle infrastructure either had no significant effect (in the case of Class II and III facilities in San Francisco) or a reduction in the likelihood of turnover for Class II and III facilities in Alameda County.

Through a survey of shoppers on matched pair corridors in San Francisco and Alameda Counties, we failed to find a coherent distinction in shopper behavior on corridors with and without bicycle infrastructure. Similar rates of shoppers arriving by bike were seen on corridors with and without bicycle infrastructure. When asked about the general frequency with which they visit the corridor by bike for shopping trips, all respondents reported coming to corridors without bicycle infrastructure more frequently. It appears that cyclists in the survey sample were wealthier on the whole than shoppers using other modes and that any differences in shopping patterns were likely due to income rather than mode used or the presence of bicycle infrastructure.

### Limitations and future research

There are a number of limitations to this research. This work was bounded in part by open data and the incompleteness of that information. The analysis focuses on only two counties (San Francisco and Alameda) rather than being able to speak comprehensively about patterns throughout the region. This is due in large part due to the lack of complete information about bicycle infrastructure in the Bay Area. The most comprehensive dataset of regional bike facilities is the MTC Regional Bike Facilities dataset. This dataset ostensibly includes the location and class of all bike infrastructure throughout the Bay Area through the year 2014. Unfortunately, it does not include installation dates. San Francisco is a model for bike infrastructure data management, publishing a comprehensive and regularly updated dataset with information on every bicycle facility in the city, including data points on the month and year of installation, facility type, and other special considerations. Contingent on

resource availability more public agencies should seek to achieve the same level of transparency. Without a reliable source for this data, installation dates were assigned manually using satellite imagery. However, the manual process is susceptible to human error, grainy historic satellite imagery, and limited imagery for years before the 2000s.

Another limitation to this research regards the intercept survey of shoppers. The overall sample size is smaller than would have been ideal, making it difficult to reliably generalize to the entire population. Future research should devote more resources to gathering a larger sample of shoppers.

In a similar vein, historical parking count data was also gathered manually using Google Street View. This process is also prone to human error, particularly when parking spaces do not have clear curb markings and/or meters. In these cases, research assistants had to make a judgment call on the number of parking spaces on a block. This is a somewhat subjective process, which introduces some irregularity into the parking count dataset. Furthermore, because of time constraints, the counts were only conducted on corridors with bicycle infrastructure. That being said, parking could have been removed for a number of other reasons on blocks without bike infrastructure, including for purposes such as parklets, bus bulbs, or transit-only lanes. The model oversimplifies the change in parking on corridors without bicycle infrastructure by assuming that these blocks saw no change in parking.

This research also doesn't touch on another issue that is likely key to business owners: the question of rents for storefront establishments. Future research should use hedonic regression to investigate whether distance to bicycle infrastructure or location immediately on bicycle infrastructure have a statistically significant association with increases in rent. A common implication in opposition to bicycle infrastructure projects from business owners and communities alike is that bicycle infrastructure projects will attract outside residents to the neighborhood, initiating or intensifying gentrification pressures (Levin 2017).

Future research should also investigate how businesses that open after the installation of bicycle infrastructure differ from businesses that were there in years prior. Characteristics such as average sales and industry group could contribute to the discussion surrounding bike infrastructure and gentrification.

A final conceptual limitation to this research is that it assumes that location immediately adjacent to bicycle infrastructure has a distinct impact from being located one block away from a corridor with bicycle infrastructure. The dichotomy between abutting and non-abutting businesses is not so black and white in the real world. Rather, a more nuanced representation of this relationship would incorporate some sort of bicycle accessibility measure to assign a weighted accessibility bonus to businesses based on their distance from bicycle infrastructure.

## Implications for planning and policy

This research explores the narrative emerging from the business community that bicycle infrastructure has a negative affect on businesses. Through an empirical analysis of secondary data and primary intercept survey data, we find a mixed association between bicycle infrastructure and business performance. Bicycle infrastructure was associated with a generally neutral effect in terms of change in sales and a positive or neutral effect in terms of turnover. These findings make it difficult to generalize bike infrastructure as being overall good or bad for the business community. Bicycle infrastructure was associated with no significant effect in many of the models produced for this report.

Models of sales change for Alameda and San Francisco Counties indicated that bicycle infrastructure alone has no significant effect on business sales, but roadway type matters. There is some evidence to support planning new bike facilities on neighborhood-type roads, but planners should carefully assess potential impacts when placing lanes on other roadway types.

The models of turnover suggest that businesses on Class II infrastructure in Alameda County could benefit from infrastructure projects in terms of a reduced likelihood of turnover. Class III facilities were also associated with a reduction in the likelihood of turnover for Alameda County. This could be interpreted as a reason to promote the expansion of bike infrastructure projects in commercial zones. With mixed findings cropping up, however, it is difficult to conclusively rule on the relationship between bike infrastructure and business performance.

Another salient issue related to bicycle infrastructure and business performance is the issue of parking removal. While only the models for San Francisco Class II facilities are able to address this issue, the results indicate no significant effect in terms of change in sales or turnover. Additionally, the presence of other curb management zones, including passenger and/or commercial loading zones, did not appear to have a significant effect. This information is also important for planners to know as they negotiate reallocations of road space for uses that conflict with auto and business owners' desire to maintain on-street parking.

In sum, the positive effects of bicycle infrastructure on bicycle mode share, cyclist safety, car usage, and population physical activity levels have been well-documented, but the same amount of attention has not been paid to the interaction between bike infrastructure and business performance. A prevailing assumption has been that business owners will oppose any change to the status quo when it comes to the road space outside their front door. While the relationship between business owners and bicycle advocates has been fraught for some time, this need not be the case. This research indicates that overall, bike infrastructure does not have a definitively positive or negative effect on business performance. Instead, there are a multitude of other factors outside of planners' control that determine sales or the likelihood that a business closes. This research can inform future conversations

around the relationship between bikes and business, and will hopefully inspire further research into this complex relationship.

## Appendix 1: List of storefront NAICS codes

NAICS Code	2012 NAICS US Title
31212	Breweries
31213	Wineries
31214	Distilleries
44111	New Car Dealers
44112	Used Car Dealers
44122	Motorcycle, Boat, and Other Motor Vehicle Dealers
44131	Automotive Parts and Accessories Stores
44132	Tire Dealers
44211	Furniture Stores
44221	Floor Covering Stores
44229	Other Home Furnishings Stores
44314	Electronics and Appliance Stores
44412	Paint and Wallpaper Stores
44413	Hardware Stores
44422	Nursery, Garden Center, and Farm Supply Stores
44511	Supermarkets and Other Grocery (except Convenience) Stores
44512	Convenience Stores
44521	Meat Markets
44522	Fish and Seafood Markets
44523	Fruit and Vegetable Markets
44529	Other Specialty Food Stores
44531	Beer, Wine, and Liquor Stores
44611	Pharmacies and Drug Stores
44612	Cosmetics, Beauty Supplies, and Perfume Stores
44613	Optical Goods Stores
44619	Other Health and Personal Care Stores
44711	Gasoline Stations with Convenience Stores
44719	Other Gasoline Stations
44811	Men's Clothing Stores
44812	Women's Clothing Stores
44813	Children's and Infants' Clothing Stores
44814	Family Clothing Stores
44815	Clothing Accessories Stores
44819	Other Clothing Stores
44821	Shoe Stores
44831	Jewelry Stores
44832	Luggage and Leather Goods Stores

45111	Sporting Goods Stores
45112	Hobby, Toy, and Game Stores
45113	Sewing, Needlework, and Piece Goods Stores
45114	Musical Instrument and Supplies Stores
45121	Book Stores and News Dealers
45211	Department Stores
45291	Warehouse Clubs and Supercenters
45299	All Other General Merchandise Stores
45311	Florists
45321	Office Supplies and Stationery Stores
45322	Gift, Novelty, and Souvenir Stores
45331	Used Merchandise Stores
45391	Pet and Pet Supplies Stores
45392	Art Dealers
45399	All Other Miscellaneous Store Retailers
51213	Motion Picture and Video Exhibition
52211	Commercial Banking
52212	Savings Institutions
52213	Credit Unions
53221	Consumer Electronics and Appliances Rental
53222	Formal Wear and Costume Rental
53223	Video Tape and Disc Rental
53229	Other Consumer Goods Rental
53231	General Rental Centers
54121	Accounting, Tax Preparation, Bookkeeping, and Payroll Services
54192	Photographic Services
54193	Translation and Interpretation Services
54194	Veterinary Services
56141	Document Preparation Services
56151	Travel Agencies
56174	Carpet and Upholstery Cleaning Services
62111	Offices of Physicians
62121	Offices of Dentists
62131	Offices of Chiropractors
62132	Offices of Optometrists
62133	Offices of Mental Health Practitioners (except Physicians)
62134	Offices of Physical, Occupational and Speech Therapists, and Audiologists
62139	Offices of All Other Health Practitioners
62141	Family Planning Centers
62142	Outpatient Mental Health and Substance Abuse Centers
62149	Other Outpatient Care Centers

71111	Theater Companies and Dinner Theaters
71112	Dance Companies
71113	Musical Groups and Artists
71119	Other Performing Arts Companies
71121	Spectator Sports
71211	Museums
71312	Amusement Arcades
71321	Casinos (except Casino Hotels)
71329	Other Gambling Industries
71394	Fitness and Recreational Sports Centers
71395	Bowling Centers
72233	Mobile Food Services
72241	Drinking Places (Alcoholic Beverages)
72251	Restaurants and Other Eating Places
81111	Automotive Mechanical and Electrical Repair and Maintenance
81112	Automotive Body, Paint, Interior, and Glass Repair
81119	Other Automotive Repair and Maintenance
81141	Home and Garden Equipment and Appliance Repair and Maintenance
81142	Reupholstery and Furniture Repair
81143	Footwear and Leather Goods Repair
81149	Other Personal and Household Goods Repair and Maintenance
81211	Hair, Nail, and Skin Care Services
81219	Other Personal Care Services
81231	Coin-Operated Laundries and Drycleaners
81232	Drycleaning and Laundry Services (except Coin-Operated)
81233	Linen and Uniform Supply
81291	Pet Care (except Veterinary) Services
81292	Photofinishing

## Appendix 2: Industry groupings

Industry	Number	Description
<b>Auto-oriented</b>	44111	New Car Dealers
	44112	Used Car Dealers
	44122	Motorcycle, Boat, and Other Motor Vehicle Deal...
	44131	Automotive Parts and Accessories Stores
	44132	Tire Dealers
	44711	Gasoline Stations with Convenience Stores
	44719	Other Gasoline Stations
	81111	Automotive Mechanical and Electrical Repair an...
	81112	Automotive Body, Paint, Interior, and Glass Re...
81119	Other Automotive Repair and Maintenance	
<b>Bar</b>	7224	Drinking Places (Alcoholic Beverages)
<b>Entertainment</b>	711	Performing Arts, Spectator Sports, and Related Industries
	712	Museums, Historical Sites, and Similar Institutions
	713	Amusement, Gambling, and Recreation Industries
	51213	Motion Picture and Video Exhibition
<b>Financial</b>	52211	Commercial Banking
	52212	Savings Institutions
	52213	Credit Unions
	52219	Other Depository Credit Intermediation
	54121	Accounting, Tax Preparation, Bookkeeping, and Payroll Services
	522291	Consumer Lending
<b>Grocery</b>	445	Grocery Stores
<b>Health</b>	621	Ambulatory Health Care Services
<b>Home goods</b>	44221	Floor Covering Stores
	44229	Other Home Furnishings Stores
	44314	Electronics and Appliance Stores
	44412	Paint and Wallpaper Stores
	44413	Hardware Stores
	44422	Nursery, Garden Center, and Farm Supply Stores
	81142	Reupholstery and Furniture Repair
443142	Electronics Stores	
<b>Personal shopping</b>	446	Health and Personal Care Stores
	448	Clothing Stores
	4511	Sporting Goods, Hobby, and Musical Instrument Stores

	4512	Book Stores and News Dealers
	4521	Department Stores
	45391	All Other Miscellaneous Store Retailers
<b>Restaurant</b>	7225	Restaurants and Other Eating Places
<b>Service</b>	8121	Personal Care Services
	8123	Drycleaning and Laundry Services
	8129	Other Personal Services
	54192	Photographic Services
	54193	Translation and Interpretation Services
	81143	Footwear and Leather Goods Repair
	81211	Hair, Nail, and Skin Care Services
	81219	Other Personal Care Services
	81231	Coin-Operated Laundries and Drycleaners
	81233	Linen and Uniform Supply
	81291	Pet Care (except Veterinary) Services
	81292	Other Personal Care Services
	811213	Communication Equipment Repair and Maintenance

## References

- Andersen, Michael, and Mary Lauran Hall. 2014. "Protected Bike Lanes Mean Business." PeopleForBikes.  
[https://b.3cdn.net/bikes/123e6305136c85cf56\\_0tm6vjeuo.pdf](https://b.3cdn.net/bikes/123e6305136c85cf56_0tm6vjeuo.pdf).
- Bent, Elizabeth, and Krute Singa. 2009. "Modal Choices and Spending Patterns of Travelers to Downtown San Francisco, California: Impacts of Congestion Pricing on Retail Trade." *Transportation Research Record: Journal of the Transportation Research Board* 2115 (December): 66–74.  
<https://doi.org/10.3141/2115-09>.
- Bialick, Aaron. 2013. "Fearmongering Overwhelms Facts at Meeting About Livable Polk Street." *Streetsblog SF* (blog). March 19, 2013.  
<https://sf.streetsblog.org/2013/03/19/fearmongering-overwhelms-facts-at-meeting-about-livable-polk-street/>.
- Boone, Andrew. 2014a. "East San Jose Bikeway Plan Scrutinized, Park Avenue Parking Debate Begins." *Streetsblog SF* (blog). August 11, 2014.  
<https://sf.streetsblog.org/2014/08/11/east-san-jose-bikeways-proposed-as-park-avenue-parking-removal-debate-begins/>.
- . 2014b. "San Jose Merchants Object to Parking Removal for Bike Lanes on Park Ave." *Streetsblog SF* (blog). August 19, 2014.  
<https://sf.streetsblog.org/2014/08/19/san-jose-merchants-object-to-parking-removal-for-bike-lanes-on-park-ave/>.
- Clifton, Kelly. 2012. "Consumer Behavior and Travel Choices: A Focus on Cyclists and Pedestrians," August.  
<http://www.ledevoir.com/documents/pdf/etudeportland.pdf>.
- "Columbus Avenue Neighborhood Transportation Study." 2010. San Francisco, CA: San Francisco County Transportation Authority.  
<http://www.sfcta.org/transportation-planning-and-studies/neighborhood-transportation-planning/columbus-avenue-neighborhood-transportation-study>.
- Crane, Melanie, Chris Rissel, Stephen Greaves, Chris Standen, and Li Ming Wen. 2016. "Neighbourhood Expectations and Engagement with New Cycling Infrastructure in Sydney, Australia: Findings from a Mixed Method before-and-after Study." *Journal of Transport & Health* 3 (1): 48–60.  
<https://doi.org/10.1016/j.jth.2015.10.003>.
- Curry, Melanie. 2016. "Oakland's New Parking Protected Bike Lanes Are Challenging to Some." *Streetsblog SF* (blog). May 4, 2016.  
<https://sf.streetsblog.org/2016/05/04/oaklands-new-parking-protected-bike-lanes-are-challenging-to-some/>.
- . 2017. "Telegraph Avenue Parking-Protected Bike Lanes Show Stunning Results." *Streetsblog California* (blog). January 30, 2017.  
<https://cal.streetsblog.org/2017/01/30/telegraph-avenue-parking-protected-bike-lanes-show-stunning-results/>.

- Dooling, Sarah. 2009. "Ecological Gentrification: A Research Agenda Exploring Justice in the City." *International Journal of Urban and Regional Research* 33 (3): 621–39. <https://doi.org/10.1111/j.1468-2427.2009.00860.x>.
- Drennen, Emily. 2003. "Economic Effects of Traffic Calming on Urban Small Businesses." Department of Public Administration, San Francisco State University. <http://bbkohyp.sfbike.org/download/bikeplan/bikelanes.pdf>.
- "East Village Shoppers Study: A Snapshot of Travel and Spending Patterns of Residents and Visitors in the East Village." 2012. Transportation Alternatives. [https://www.transalt.org/sites/default/files/news/reports/2012/EV\\_Shopper\\_Study.pdf](https://www.transalt.org/sites/default/files/news/reports/2012/EV_Shopper_Study.pdf).
- Esper, Damin. 2013. "Albany Adopts Plan for San Pablo Avenue." *East Bay Times*, December 18, 2013. <https://www.eastbaytimes.com/2013/12/18/albany-adopts-plan-for-san-pablo-avenue/>.
- Fine, Sarah. 2017. "Telegraph Avenue Progress Report." Oakland, CA: Oakland DOT. <http://www2.oaklandnet.com/oakca1/groups/pwa/documents/report/oak062598.pdf>.
- Flusche, Darren. 2012. "Bicycling Means Business: The Economic Benefits of Bicycle Infrastructure." League of American Bicyclists. [https://bikeleague.org/sites/default/files/Bicycling\\_and\\_the\\_Economy-Econ\\_Impact\\_Studies\\_web.pdf](https://bikeleague.org/sites/default/files/Bicycling_and_the_Economy-Econ_Impact_Studies_web.pdf).
- Forkes, Jennifer, and N Smith-Lea. 2010. "Bike Lanes, On-Street Parking, and Business, Year 2 Report: A Study of Bloor Street in Toronto's Bloor West Village." Clean Air Partnership.
- Gould, Kenneth Alan, and Tammy L. Lewis. 2017. *Green Gentrification: Urban Sustainability and the Struggle for Environmental Justice*. Abingdon, Oxon ; New York, NY: Routledge.
- Kroll, Cynthia A., Diana Lee, and Nadir Shams. 2010. "The Dot-Com Boom and Bust in the Context of Regional and Sectoral Changes." *Industry & Innovation* 17 (1): 49–69. <https://doi.org/10.1080/13662710903573836>.
- Krueger, R, and Gibbs, eds. 2007. "The Sustainable Development Paradox: Urban Political Economy in the United States and Europe." <https://doi.org/10.5840/enviroethics200931337>.
- Lee, Allison. 2008. "What Is the Economic Contribution of Cyclists Compared to Car Drivers in Inner Suburban Melbourne's Shopping Strips?" Master's Thesis, Melbourne, Australia: University of Melbourne.
- Levin, Sam. 2017. "'It's Not for Me': How San Francisco's Bike-Share Scheme Became a Symbol of Gentrification." *The Guardian*, August 21, 2017. <https://www.theguardian.com/us-news/2017/aug/21/bike-sharing-scheme-san-francisco-gentrification-vandalism>.
- Lubitow, Amy, Bryan Zinschlag, and Nathan Rochester. 2016. "Plans for Pavement or for People? The Politics of Bike Lanes on the 'Paseo Boricua' in Chicago, Illinois." *Urban Studies* 53 (12): 2637–53. <https://doi.org/10.1177/0042098015592823>.
- McCann, B, and S Rynne. 2010. "Complete Streets: Best Policy and Implementation Practices." American Planning Association.

- “Measuring the Street: New Metrics for 21st Century Streets.” 2012. New York, NY: NYC DOT. <http://www.nyc.gov/html/dot/downloads/pdf/2012-10-measuring-the-street.pdf>.
- Moore, Derek. 2011. “Sonoma Commotion: Bikes vs. Parking.” *The Press Democrat*, January 17, 2011. <http://www.pressdemocrat.com/news/2288039-181/sonoma-commotion-bikes-vs-parking>.
- Neumark, David, Junfu Zhang, and Brandon Wall. 2007. “Employment Dynamics and Business Relocation: New Evidence from the National Establishment Time Series.” In *Research in Labor Economics*, 26:39–83. Bingley: Emerald (MCB UP ). [https://doi.org/10.1016/S0147-9121\(06\)26002-3](https://doi.org/10.1016/S0147-9121(06)26002-3).
- O’Connor, David, James Nix, Simon Bradshaw, and Enda Shiel. 2011. “Shopping Travel Behaviour in Dublin City Centre.” ITRN2011. Cork: University College Cork. <http://arrow.dit.ie/comlinkoth/10/>.
- Poirier, Joseph. 2018. “Bicycle Lanes and Business Success: A San Francisco Examination.” *Transportation Research Record: Journal of the Transportation Research Board* (forthcoming).
- Popovich, Natalie, and Susan Handy. 2014. “Bicyclists as Consumers: Mode Choice and Spending Behavior in Downtown Davis, California.” *Transportation Research Record: Journal of the Transportation Research Board* 2468 (December): 47–54. <https://doi.org/10.3141/2468-06>.
- . 2015. “Downtown, Strip Centers, and Big-Box Stores: Mode Choice by Shopping Destination Type in Davis, California.” *Journal of Transport and Land Use* 8 (2). <https://doi.org/10.5198/jtlu.2015.739>.
- Quastel, Noah. 2009. “Political Ecologies of Gentrification.” *Urban Geography* 30 (7): 694–725. <https://doi.org/10.2747/0272-3638.30.7.694>.
- Raco, Mike. 2005. “Sustainable Development, Rolled-out Neoliberalism and Sustainable Communities.” *Antipode* 37 (2): 324–47. <https://doi.org/10.1111/j.0066-4812.2005.00495.x>.
- Ramm, Frederik. 2017. “OpenStreetMap Data in Layered GIS Format.” GEofabrik. <https://www.geofabrik.de/data/geofabrik-osm-gis-standard-0.7.pdf>.
- Rankin, Katharine N., and Heather McLean. 2015. “Governing the Commercial Streets of the City: New Terrains of Disinvestment and Gentrification in Toronto’s Inner Suburbs: Governing the Commercial Streets of the City.” *Antipode* 47 (1): 216–39. <https://doi.org/10.1111/anti.12096>.
- Rowe, Kyle. 2013. “Bikenomics: Measuring the Economic Impact of Bicycle Facilities on Neighborhood Business Districts.” Seattle, Washington: University of Washington College of Built Environments. [http://bikewalkkc.org/wp-content/uploads/2016/03/Bikenomics\\_v4.pdf](http://bikewalkkc.org/wp-content/uploads/2016/03/Bikenomics_v4.pdf).
- Rudick, Roger. 2017. “Valencia Human-Protected Bike Lane Protest #2.” *Streetsblog SF* (blog). May 26, 2017. <https://sf.streetsblog.org/2017/05/26/valencia-human-protected-bike-lane-protest-2/>.
- San Francisco Bicycle Coalition. 2017. “Polk Street.” 2017. <http://www.sfbike.org/our-work/street-campaigns/polk-street/>.
- SFMTA. 2017. “Polk Streetscape Project.” 2017. <https://www.sfmta.com/projects/polk-streetscape-project>.

- Shaer, Matthew. 2011. "Not Quite Copenhagen: Is New York Too New York for Bike Lanes?" *New York Magazine*, March 20, 2011, sec. Features.  
<http://nymag.com/news/features/bike-wars-2011-3/>.
- Siemiatycki, Matti, Matt Smith, and Alan Walks. 2016. "The Politics of Bicycle Lane Implementation: The Case of Vancouver's Burrard Street Bridge." *International Journal of Sustainable Transportation* 10 (3): 225–35.  
<https://doi.org/10.1080/15568318.2014.890767>.
- Stantec Consulting Ltd. 2011. "Vancouver Separated Bike Lane Business Impact Study." Vancouver, BC.  
<http://council.vancouver.ca/20110728/documents/penv3-BusinessImpactStudyReportDowntownSeparatedBicycleLanes-StantecReport.pdf>.
- Stehlin, John. 2015. "Cycles of Investment: Bicycle Infrastructure, Gentrification, and the Restructuring of the San Francisco Bay Area." *Environment and Planning A* 47 (1): 121–37. <https://doi.org/10.1068/a130098p>.
- Szczepanski, Carolyn. 2013. "How Bicycles Bring Business." *Momentum Mag* (blog). April 10, 2013. <https://momentummag.com/how-bicycles-bring-business/>.
- Sztabinski, Fred. 2009. "Bike Lanes, on-Street Parking and Business: A Study of Bloor Street in Toronto's Annex Neighborhood."
- Tolley, Rodney. 2011. "Good for Busine\$\$ - The Benefits of Making Streets More Walking And Cycling Friendly." Heart Foundation South Australia.  
[www.heartfoundation.org.au/SiteCollectionDocuments/GoodforBusinessFINAL\\_Nov.pdf](http://www.heartfoundation.org.au/SiteCollectionDocuments/GoodforBusinessFINAL_Nov.pdf).
- Vreugdenhil, Roger, and Stewart Williams. 2013. "White Line Fever: A Sociotechnical Perspective on the Contested Implementation of an Urban Bike Lane Network: White Line Fever." *Area* 45 (3): 283–91.  
<https://doi.org/10.1111/area.12029>.
- Walls & Associates. 2017. "NETS Database Description."
- Yen, Barbara, W Tseng, and M Ghafoor. 2015. "Do Restaurant Precincts Need More Parking? Differences in Business Perceptions and Customer Travel Behaviour in Brisbane, Queensland, Australia." In . Sydney, Australia.
- Zavestoski, S, and J Ageyeman, eds. 2014. *Incomplete Streets: Processes, Practices, and Possibilities*. Abingdon, Oxon: Routledge.