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How the Built Environment Affects Car Ownership and Travel: Evidence from San Francisco Housing Lotteries

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How the Built Environment Affects Car Ownership and Travel: Evidence from San Francisco Housing Lotteries

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# Technical Report Documentation Page

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How the Built Environment Affects Car Ownership and Travel: Evidence from San Francisco Housing Lotteries

Adam Millard-Ball* Jeremy West Nazanin Rezaei Garima Desai
University of California, Santa Cruz
July 2020

Abstract
Credibly identifying how the built environment shapes behavior is empirically challenging, because people select residential locations based on differing constraints and preferences for site amenities. Our study overcomes these research barriers by leveraging San Francisco’s affordable housing lotteries, which randomly allow specific households to move to specific residences. Using administrative data, we demonstrate that lottery-winning households’ baseline preferences are uncorrelated with their allotted residential features such as public transportation accessibility, parking availability, and bicycle infrastructure—meaning that neighborhood attributes and a building’s parking supply are effectively assigned at random. Surveying the households, we find that these attributes significantly affect transportation mode choices. Most notably, we show that essentially random variation in on-site parking availability greatly changes households’ car ownership decisions and driving frequency, with substitution away from public transit. In contrast, we find that parking availability does not affect employment or job mobility. Overall, the evidence from our study robustly supports that local features of the built environment are important determinants of transportation behavior.

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Executive Summary

Cities worldwide are increasingly promoting walkable, transit-oriented development as a means to reduce traffic congestion, greenhouse gas emissions, and air pollution, and improve the quality of life for their residents. Behind these efforts lies a large body of research that suggests that public transportation access, walkability, parking provision, and other physical characteristics of the built environment affect people’s car ownership, travel behavior, carbon footprints, and employment opportunities. However, households choose where to live in part based on local transportation options and parking provision. Those who prefer to ride the bus, for example, are likely to seek out housing near to frequent transit service, while households who want to own a car will gravitate towards residential buildings with on-site parking. Because of these self-selection and sorting biases, much of the current research on land use and travel behavior may be unreliable.

Our study overcomes these biases by using a survey of 779 successful applicants to San Francisco’s affordable housing lotteries to provide causal estimates of how neighborhood characteristics affect travel and employment. We demonstrate that people are not selective in which lotteries they enter—not surprisingly, given that the average success rate is 1.2 percent—meaning that lottery winners’ neighborhood attributes and residential parking supply are effectively assigned at random. Our survey shows that these factors significantly affect transportation mode choices. Even small improvements in local transit accessibility substantially increase peoples’ decisions to use transit rather than driving. Most notably, we show that providing more on-site parking increases households’ propensity to own and use a car, while decreasing use of public transit. Both on-site parking and local transit accessibility affect households’ frequency of public transit use, but the effect of parking is approximately three times as large. A natural concern about reducing residential parking is that this might limit access to jobs, but we find no evidence of this trade-off—on-site residential parking has no detectable impact on employment or job mobility.

Overall, the evidence from our study robustly supports the conclusion that local features of the built environment, particularly parking, are important determinants of transportation behavior. Moreover, reducing space dedicated to parking appears to come without employment downsides. Our findings support the efforts of many cities to remove minimum parking requirements from zoning codes and possibly replace them with maxima instead. Such zoning reforms could also yield other benefits including reducing housing costs and increasing land available for new housing and commercial development, as well as reducing motor vehicle trips and associated harms. Where streets are relatively walkable and transit service is frequent, parking emerges as the key factor shaping household travel behavior. Encouragingly, parking is a factor that is highly amenable to low-cost policy reforms that can rapidly provide significant benefits.
1 Introduction

A person’s residential neighborhood shapes their transportation habits, health, and employment —indeed, almost every aspect of their lifestyle, identity, and opportunities. In turn, the choices people make based on residential location also affect others, through externalities such as pollution, road congestion, and traffic collisions. Thus, urban planners and policymakers increasingly face calls to promote walkability, raise allowable building heights and densities, and reduce the amount of space dedicated to automobile parking. In principle, policies that provide more flexibility for developers will promote a mixture of local amenities and infrastructure that better matches the preferences of the community and allow more households to move to their preferred locations, thereby reducing the implicit regulatory tax imposed by many zoning regulations (Levine, 2005; Glaeser and Gyourko, 2018). In practice, the efficacy of these land-use policies in reducing transportation-related externalities depends heavily on how features of the built environment ultimately affect people’s behavior.

A voluminous literature in urban planning and economics considers how neighborhood attributes such as public transportation access, residential density, and walkability relate to automobile ownership, vehicle miles traveled, and emissions (e.g. Ewing and Cervero, 2010; Salon et al., 2012; Stevens, 2017). To a lesser extent, researchers have also investigated how the proximity of job opportunities and their accessibility by transit or private car correlate with employment and household income (Sanchez, 1999; Marinescu and Rathelot, 2018). Several recent studies additionally examine how residential parking availability may change a household’s propensity to own a car and how much to drive, as well as the relationship between the income mix within a neighborhood or building and residents’ social interactions and earnings (Kleit, 2005; Galster et al., 2008; Weinberger, 2012).

A significant challenge for understanding how location-based amenities such as public transportation affect residents’ travel behavior and employment opportunities, however, is that people choose where to live, and they do so based in part on local factors such as the availability of parking and public transportation. This self-selection into residential (and workplace) locations means that the vast majority of inferences from the transportation-land use literature are susceptible to selection bias (van Wee, 2009).1 Fundamentally, the empirical concern is that residential location is a decision made by the residents, rather than an assignment based on some external process. For instance, individuals who do not own cars or otherwise prefer to commute via public tran-

1Self-selection is also a major concern in the broader literature on neighborhood effects. For example, non-random residential sorting typically confounds attempts to identify how factors such as racial segregation and pollution impact social and economic outcomes (e.g. Graham, 2018; Christensen et al., 2020).
sit are more likely to try to live nearby to major rail or bus lines, biasing upward any observed correlation between transit access and utilization (Glaeser et al., 2008). Individuals who prefer owning cars and driving, on the other hand, will care more about the provision of parking. Thus, most estimates in the literature that relate infrastructure features to travel behavior lack a causal interpretation (Sampson et al., 2002).

For some policy purposes, such as predicting the volume of transit ridership, residential self-selection may be a relatively muted concern (Chatman, 2014). Often, however, a causal interpretation is desired, particularly when seeking to understand how policy-induced changes to the built environment will affect travel behavior and transportation mode decisions (Ewing and Cervero, 2010). For instance, planners setting municipal requirements for residential parking can better serve their communities if they know how parking availability actually affects local automobile use.

Studies in this literature often attempt to correct for residential selection bias using statistical controls, instrumental variables, or related methods (see Cao et al., 2009, for a review). These bias-correcting techniques considerably change quantitative results, including reducing the estimated impacts of land-use characteristics like urban density on vehicle travel by fifty percent or more (Stevens, 2017). However, structural models and quasi-experimental approaches require strong assumptions and are only partial solutions, especially considering that the selection bias involves both observable and unobservable factors (Pinjari et al., 2007). In short, the empirical challenge is that residential “self-selection leads to non-random heterogeneity in choices and behaviour” (van Wee, 2009).

As in nearly all areas of social science research, randomized experiments are the gold standard to identify causal effects. In principle, researchers could randomly assign households to different types of neighborhoods and then observe their behavior, but this is rarely practical or ethical (Cao et al., 2009). Randomization has been successfully employed to analyze how federally-subsidized housing vouchers via the Moving to Opportunity program affect economic opportunities, crime, and public health outcomes (Katz et al., 2001; Ludwig et al., 2001; Leventhal and Brooks-Gunn, 2003; Kling et al., 2007; Sanbonmatsu et al., 2011; Chetty et al., 2016). Similar lotteries for housing assistance or public housing have also been used to analyze labor market and health outcomes in Canada, Ethiopia, India, and the Netherlands (Adair et al., 2016; Barnhardt et al., 2017; Bowen et al., 2018; Franklin, 2019; van Dijk, 2019). All of these studies, however, primarily evaluate the effects of randomly moving households away from particular residential locations—such as out of government-provided housing projects—rather than the effects of assigning people to live in specific residential locations.
In this report, we leverage the housing lottery programs in San Francisco to overcome the aforementioned research limitations and provide causal interpretations of the impacts of specific neighborhood characteristics and parking provision on households’ transportation behavior and economic outcomes. In San Francisco, nearly all new housing developments with ten or more residential units must offer a government-specified share of “inclusionary” units at below-market-rate (BMR) prices, either directly on-site, directly off-site, or indirectly off-site by paying a fee. As might be expected, demand for new BMR units substantially exceeds the available supply—one recent lottery for 95 rental units attracted 6,580 household applicants (Badger, 2018). Because of the very low odds of winning, eligible households generally apply indiscriminately to many different housing lotteries. Those that are fortunate to eventually win a BMR unit are thus effectively assigned to live in specific buildings and neighborhoods. In essence, San Francisco’s housing lotteries provide as-good-as-random assignment of people into homes.

Our research design compares transportation behavior and economic outcomes across households that won different BMR lotteries, which thereby provides effectively random variation in their residential building characteristics like on-site parking availability and neighborhood-level characteristics such as bicycle infrastructure and accessibility of public transportation. In doing so, we provide the first evidence to our knowledge about transportation behavior and economic outcomes for a population that is in effect randomly assigned to live in particular places. To validate our empirical strategy, we assess whether households are selective in the types of BMR housing projects for which they apply, finding that lottery participation decisions are indeed as-good-as-random. In other words, by studying a sample of lottery winners we eliminate the possibility of self-selection bias; these households were no more or less likely to move to particular neighborhoods because of the availability of transit or parking. We present findings from a survey that we conducted of about 2,700 of these households currently residing in BMR units, asking them questions about their transportation choices and employment.

The responses to our survey confirm the importance of accessibility by walking, bicycling, and transit in shaping household transportation choices. Even in a city such as San Francisco, where walkable neighborhoods are the norm and public transit quality is quite high by U.S. standards, accessibility substantially impacts people’s travel and commuting decisions. On-site residential parking has even larger effects: increased parking causes more car ownership and more driving while reducing transit use, regardless of a neighborhood’s transit accessibility. Moreover, additional parking does not improve employment or labor market mobility among households in our sample. In summary, the evidence from our study robustly supports that urban residents’ transportation behavior—but not their employment—is affected by local features of the built environ-
ment, and particularly so by parking.

The remainder of this report is organized as follows. First, we provide some context for the below-market-rate housing programs in San Francisco. We then discuss our empirical methodology, survey design, and data. Next, we validate our assumption that housing assignment is as-good-as random by examining household lottery participation behavior. Finally, we present the findings from our survey of travel behavior and employment outcomes. We conclude by discussing policy implications for affordable housing and urban parking requirements.

2 Setting: San Francisco affordable housing programs

San Francisco is often ranked as one of the least affordable cities in the United States (e.g. NAHB, 2019). In response, the city has developed and implemented a range of programs to increase the availability of affordable housing and to provide down payment assistance for qualified home purchases. Most of these programs are administered by the Mayor’s Office of Housing and Community Development (MOHCD).\(^2\) In this report, we focus on the Inclusionary Housing program, under which a government-specified portion of units in most new residential developments must be made available at below-market-rate prices (or rent) to households whose income is below specified thresholds. Given San Francisco’s high incomes, a household can qualify while earning up to $118,200 for a two-person household, equivalent to 120 percent of area median income.

The BMR percentage requirements and thresholds have varied over time since the Inclusionary Housing program was established in 2002, but as of December 2019, twelve to twenty percent of on-site housing units must be set aside for low- or middle-income households. Alternatively, developers can directly provide off-site affordable housing or pay a fee that is used to supply of off-site affordable housing.\(^3\)

Inclusionary housing is part of a range of affordable housing programs administered by MOHCD and other city and nonprofit agencies. BMR housing projects are also developed using a mix of public and private funds by the city’s Office of Community Investment and Infrastructure, which is the successor to the city’s Redevelopment Agency. Multifamily projects, meanwhile, typically consist fully of affordable units and are often developed by non-profit organizations. Some projects cater to specific groups such as seniors, people who are homeless, and people with disabilities.

\(^2\)San Francisco’s program details including eligibility criteria and funding are available at sfmohcd.org.

\(^3\)See https://sfplanning.org/project/inclusionary-affordable-housing-program for details. In general, a twelve percent requirement applies to properties of 10-24 units; eighteen percent to larger (25+ unit) rental properties; and twenty percent to larger ownership properties. Some neighborhoods have higher requirements. Smaller properties of fewer than ten units are exempt.
Although specific eligibility requirements and funding sources vary between and within these programs, a lottery mechanism is used to allocate units in all of the BMR projects that we study in this report. First, would-be residents submit an electronic or paper application form for each lottery; no fee is required at this stage. Second, applicants are randomly assigned a rank. Then, eligibility is verified for those receiving a sufficiently high rank. Finally, units are offered to eligible applicants in rank order within certain priority groups.

The order of these priority groups varies by project, but in recent years a typical inclusionary housing project offers units sequentially to (i) Certificate of Preference holders, who were displaced by urban renewal in the 1960s (statutory up to 100 percent of units); (ii) tenants displaced by no-fault eviction or fire (up to 20 percent of units); (iii) neighborhood residents (up to 40 percent of units), and (iv) residents who live or work in San Francisco (up to 100 percent of units). In practice, Certificate of Preference holders and displaced tenants are few in number, and about two-thirds of successful applicants are in the live-work priority group. Thus, as we demonstrate below, these lotteries provide effectively random assignment to the vast majority of successful applicants, most of whom live or work in the neighborhood or in other parts of San Francisco.

Projects also vary by the amount of parking that is provided. In the early years of the BMR program, projects had a one-to-one ratio of parking spaces to units, and the cost of parking was bundled in with the rent or sales price. In line with subsequent zoning reforms, however, more recent projects have unbundled parking from the rent or sales price—i.e., residents are free to decline a parking space, but accepting it entails an added cost. At the same time, parking ratios of less than one space per unit or even zero parking have become more common. For projects that have a parking ratio of less than one space per unit, spaces are offered in lottery rank order within each priority group. For example, in a project with ten BMR units and a parking ratio of 0.5:1, the first five lottery winners would be guaranteed an offer of parking, but remaining lottery winners would only be offered parking if higher-ranked applicants declined to take (and normally pay for) a parking space.\footnote{Indeed, parking spaces intended for below-market-rate units often go unclaimed, even in buildings with less than one space per unit. While winning the affordable housing lottery is highly prized, households seem to care less about winning the “parking lottery.”} Developers are required to assign or offer parking spaces to BMR units at the same ratio as they provide for market-rate units.
3 Methods and Data

3.1 Housing lottery applications

To validate our assumption that housing assignments are as good as random, we use a dataset of all 107,310 applications to 59 BMR housing lotteries held between July 2015 and June 2018, which we call our applicant sample. Because the applicant sample is only available for a three-year period, it excludes many housing projects that are included in our primary survey sample. However, the applicant sample also encompasses 14 projects which are not present in our survey sample; these projects are currently managed by a nonprofit housing organization or by another city agency, precluding survey distribution to these units. In addition to lottery rank and priority group status, the applicant sample provides basic demographic information from lottery applications such as income, gender, and race.

The applicant sample consists of individual applications and does not link repeat applicants across lotteries for different projects. Therefore, we first match people who apply to multiple lotteries based on (i) their date of birth and (ii) any of the following: first name, last name, or address. We also match applicants based on all of the following: first name, last name, and address.6

3.2 Household survey design

Our primary data survey sample consists of all BMR units for which we have occupancy and parking data, and comprises 2,654 units in 197 projects that were occupied as of April 2019. Almost all (2,605) of these units were built under the Inclusionary Housing program. We obtained data on project-level characteristics directly from MOHCD and supplemented these data using land use permit approval records to fill in missing data such as parking ratios.7 As shown in Figure 1 and Table 1, our survey sample provides meaningful variation in households’ building-level and neighborhood characteristics. BMR units are distributed throughout the city, giving a range of walking environments and proximity to public transit, as well as substantial variation in on-site parking

5The application sample period is shorter due to changes in how MOHCD processes and retains lottery data.
6We use a chained matching process that iteratively links groups of applicants who are matched on each of the combinations. For example, a group that matches on first name and date of birth would be combined with another group that matches on address and date of birth, if the groups have overlapping members. The chained process, rather than relying on matching specific fields or fields is necessary because of errors, spelling variations, and missing data in the applicant dataset. Identifying information was hashed (scrambled) by MOHCD prior to providing us the dataset, in order to safeguard individual privacy. Thus, we are unable to clean errors and spelling variations manually.7The MOHCD project-level data are available at https://data.sfgov.org/Housing-and-Buildings/Affordable-Housing-Pipeline/aaxw-2cb8.
availability. Units are roughly evenly split between rentals (53 percent of units) and for-sale units; although more projects are ownership (65 percent), these tend to be smaller in scale.

Figure 1: Locations of surveyed below-market-rate residential projects in San Francisco

Source: Authors’ analysis of Mayor’s Office of Housing and Community Development (MOHCD) data.

Our survey asked all BMR residents in our survey sample about their frequency of travel by mode; car ownership; employment status; the location of the respondent’s workplace or school (if any); and their interactions with neighbors.8 These survey questions are not intended to calculate vehicle miles traveled or other common metrics of transportation usage, which would require a substantially more complex survey instrument, impacting response rates and increasing recall

8We impute some missing responses for Question 1: where a respondent left one transportation mode frequency blank but answered for other modes, we impute a response of “less often”—the lowest-frequency category.
bias. Similarly, our employment questions allow us to create coarse measures of labor market mobility. We ask about interactions with neighbors to measure one aspect of social capital.

The survey questionnaire was mailed in June 2019 to 2,654 primary occupants of inclusionary housing units in the MOHCD database. Figure 2 shows the paper survey instrument. In addition, 1,693 of these (same) occupants received an email version of the survey with a personalized link to an online survey platform with questions available in English, Spanish, Filipino and Chinese. The paper survey instrument included a multilingual URL for the online survey, as an alternative to returning the form in the reply-paid envelope. As an incentive for participating, respondents were eligible to win one of ten $100 Visa gift cards, which we awarded randomly to ten respondents. After merging cases where we received both an online and a mail-back response, we obtained 779 responses, a response rate of 29.4 percent. We attribute this high response rate to our efforts to keep the survey instrument very short (one side of an A9-size card), simple questions that are minimally intrusive, the pecuniary incentives, and the twin modes of distribution.

### 3.3 Transportation accessibility measures

Our analyses consider how four primary measures of transportation accessibility affect household behavior. We quantify private automobile accessibility using each building’s ratio of parking spaces per residential unit. We use the Center for Neighborhood Technology’s AllTransit performance
1. How often do you use the following for transportation in San Francisco?

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<td>Drive alone</td>
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<td>Carpool</td>
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<td>Public transit</td>
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<td>Biking</td>
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2. How many cars does your household own?

3. Do you work or attend school?

4. How long have you been on your current work or school?

5. What is the ZIP code or neighborhood of your current and previous work or school?

6. How do you usually get to work or school?

7. Did you talk to a neighbor in the last week?

You will be entered into a raffle to win one of ten $10 Amazon gift cards.

Thank you for completing the survey!

Notes: Figure 2 shows a postcard of the authors' survey, which was also provided online via Qualtrics to facilitate data collection.
score to measure transit frequency and quality, and we use the WalkScore company’s Walk Score and Bike Score metrics to measure accessibility by walking and cycling, respectively.

Whereas many analyses of how land-use relates to transportation behavior focus on density, land-use mix, street connectivity, and similar variables (see Ewing and Cervero, 2010, for a review), our accessibility variables arguably provide a better measure of household transportation choices in our setting. For one, each of our measures is specific to a particular mode of transportation. In contrast, factors like urban density and street connectivity can affect household decisions through multiple channels, such as by making frequent public transportation service feasible and by providing more direct travel paths for walking to local destinations (Barrington-Leigh and Millard-Ball, 2020). Accessibility is also the more proximate influence—households do not make travel decisions directly in response to density, land-use mix, or connectivity, but in response to how these factors affect accessibility. Finally, transit accessibility can be changed more directly through policy, for example by changing service frequencies or routes. We therefore focus on accessibility-oriented metrics rather than (for example) sheer density, because these proxies more comprehensively reflect the variation in households’ abilities to access destinations using particular travel modes.

As would be true for nearly any measures of transportation accessibility, our explanatory variables are correlated through spatial variation. For instance, a location that has a relatively higher Walk Score is also likely to have a relatively higher transit score, and buildings with good transit accessibility tend to have less parking. To address this collinearity, our preferred regression specifications include the parking ratio and only one of the other accessibility measures, an approach that captures the meaningful spatial variation in accessibility while providing regression estimates that can readily be interpreted.

4 Results

4.1 Demonstrating as-good-as-random housing assignment

We begin our empirical analysis by demonstrating that assignment of lottery-winning households to housing units is as-good-as-random, which facilitates causal inference. To do so, we examine the patterns of lottery participation and repeat-entrant behavior among households in our applicant sample. While each lottery is itself random by design, households might possibly choose

9In our survey sample, the correlation between a building’s AllTransit performance score and Walk Score is 0.76; the correlation between a building’s parking ratio and AllTransit performance score is -0.33.
to selectively enter only certain lotteries, for example by forgoing the chance to rent or buy in a building without parking or in one that is distant from a transit stop.

A reasonable hypothesis is that households are not selective, given the low probability of winning any single lottery. As shown in Figure 3, BMR projects attract up to 6,575 applicants, while the average lottery in our applicant sample offers only 27 units (median = 11 units; maximum = 170 units). With the exception of a handful of projects that cater to specific populations such as seniors, the odds of winning a rental unit lottery are extremely small—the average success rate of these applications is only 1.2 percent. Lotteries for ownership units attract a much smaller pool of applicants, likely because of the need to obtain a mortgage down payment, but these lotteries still have an average success rate of only 12 percent. Across all lotteries included in our applicant sample, the average success rate is 1.5 percent.\textsuperscript{10}

We confirm our hypothesis that households are not selective using regression analysis. Specifically, we estimate whether a household is more (or less) likely to participate in a particular housing lottery depending on how the characteristics of that lottery differ from those of the first lottery that the same household entered. For each household in the applicant sample, we defined the set of relevant lotteries as those with a lottery date between that of the first and last lotteries in our dataset that the household entered. We then estimate whether a household’s decision to skip or enter each of these lotteries is explained by the characteristics of the associated project. We measure project characteristics (such as the parking ratio) in terms of their absolute differences from that of the first lottery in our dataset that the household played, which we take to be the baseline preferences of the household.\textsuperscript{11}

We find no evidence that households skip lotteries based on project or neighborhood characteristics such as parking and walkability. Table 2 presents linear probability models for lottery skipping using different subsamples of lottery-applying households. The first column includes all households that we observe playing at least two lotteries (as there cannot be skipping by households that played only a single lottery). If households were selectively participating based on their baseline preferences, then we should see that an absolute change in project characteristics—relative to those of the first lottery entered by that same household—would be associated with

\textsuperscript{10}As noted above in Section 2, many applicants have some “preference” in the stratified lotteries, most often because they already live or work within San Francisco. However, even households with a “Live-Work preference”—two-thirds of successful applicants—have an average lottery success rate of only 1.7 percent. Current neighborhood residents, who are given even more priority, have an average success rate of 2.6 percent.

\textsuperscript{11}All explanatory terms are first standardized using a z-transformation (mean = 0, sd = 1). We control for household-specific fixed effects, as some households are more attentive in general to lottery availability and participate more often overall. We also control for lottery-specific fixed effects, as some lotteries are relatively better-advertised, occur at a more fortuitous time of year, have less restrictive eligibility criteria, or otherwise attract entry from a broader section of the population.
Figure 3: Number of entrants across BMR residential lotteries by application outcomes

Source: Authors’ analysis of MOHCD data.
Notes: Available application data include lotteries held from July 2015 through June 2018. Each stacked bar shows outcomes for a specific BMR residential lottery, ordered horizontally by date. “Other” lottery outcomes include applications that were withdrawn, disqualified, or that have an unknown outcome.
a larger propensity to skip a particular lottery. In contrast, the regression estimates indicate no
evidence of lottery selectivity. For instance, we find that a one standard deviation difference in
the parking ratio of the building, equivalent to 0.43 parking spaces per unit, is associated with a
tiny 0.5 percentage point decrease in the probability of the household participating in the lottery.
We find similarly small and almost always statistically insignificant relationships for the other ex-
planatory terms. The one significant estimate, for Walk Score, is quantitatively small and of the
wrong sign if participation were selective.

Table 2: Identification tests for lottery skipping: Regression estimates

<table>
<thead>
<tr>
<th>Dependent variable: I{skipped lottery}</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Abs(Δ) in std. parking ratio</td>
<td>0.005</td>
<td>0.003</td>
<td>0.010</td>
</tr>
<tr>
<td></td>
<td>(0.006)</td>
<td>(0.008)</td>
<td>(0.015)</td>
</tr>
<tr>
<td>Abs(Δ) in std. Transit Score</td>
<td>−0.001</td>
<td>−0.007</td>
<td>0.011</td>
</tr>
<tr>
<td></td>
<td>(0.005)</td>
<td>(0.017)</td>
<td>(0.023)</td>
</tr>
<tr>
<td>Abs(Δ) in std. Walk Score</td>
<td>0.014**</td>
<td>0.021</td>
<td>0.034</td>
</tr>
<tr>
<td></td>
<td>(0.005)</td>
<td>(0.014)</td>
<td>(0.020)</td>
</tr>
<tr>
<td>Abs(Δ) in std. Bike Score</td>
<td>−0.004</td>
<td>0.003</td>
<td>−0.004</td>
</tr>
<tr>
<td></td>
<td>(0.005)</td>
<td>(0.011)</td>
<td>(0.016)</td>
</tr>
</tbody>
</table>

Sample Average skip rate Household fixed effects Lottery fixed effects Number of households Number of lotteries Observations

All applicants Winner occupants Survey respondents

0.779 0.759 0.743
Yes Yes Yes
Yes Yes Yes
18,574 481 159
59 59 59
290,085 10,165 3,441

Notes: Standard errors two-way clustered by household and lottery. An ob-
servation is either a household application to a lottery (107,310 in total) or a household skipping of a lottery (182,775). Each column presents estimates from a separate regression model.

The same null patterns continue to hold as we restrict the estimation samples in Column (2) to
use only households that (eventually) won a lottery, i.e. those that we surveyed, or even further restrict in Column (3) to our survey respondents. On the whole, the evidence in Table 2 clearly supports that households—quite understandably—are not selective in their participation in these
low-odds housing lotteries.\textsuperscript{12}

Another potential source of bias is differential patterns of survey response—for example, if households in buildings with better transit access or lower parking ratios are more or less likely to respond to the survey. Table 3 uses a linear probability model to evaluate whether households’ propensities to respond to our survey vary with their project characteristics, namely, the local transportation accessibility. As above, all explanatory variables are z-standardized. Thus, for instance, the interpretation of the first coefficient in Column (4) is that, controlling for the local Walk Score, Bike Score, Transit Score, and the survey recipient’s income and demographic variables, a one standard deviation increase in the building’s parking ratio causes a 0.3 percentage point increase in the likelihood of survey response—a minuscule association. The estimates for the other independent terms likewise support that there is no survey response selection bias, as do the less-saturated specifications in Columns (1) through (3). Overall, these null results strengthen our confidence that possible bias in survey response rates is unlikely to be a factor for our subsequent findings.

4.2 Survey analysis for transportation

Having demonstrated as-good-as-random assignment of people into homes, the remainder of our analysis focuses on the household survey that we fielded. We begin by examining the relationship between household car ownership and a building’s parking provision and neighborhood transportation accessibility. Figure 4 demonstrates a clear and substantive trend: the more parking in a building, the more likely a resident household is to own a car. In buildings with no on-site parking, only 38 percent of households own a car. In buildings with at least one parking space per unit, more than 81 percent of households own automobiles. Moreover, for buildings with intermediate amounts of parking, the pattern in Figure 4 shows monotonically increasing car ownership rates.

A similar relationship between parking provision and car ownership is reflected by the regression models in Table 4. In Column (1), a minimal univariate specification indicates that a one standard deviation increase in a building’s parking ratio—about 0.43 additional spaces per unit—causes a household to be 14 percentage points more likely to own a car. As discussed above in Section 3.3, parking ratios are correlated with the other neighborhood-level factors such as transit-accessibility and walkability. However, Columns (2) through (4) show very similar estimates (12 percentage points) using specifications that also include regressors for accessibility by transit.

\textsuperscript{12}These null results are similarly unchanged when further restricting the sample to include only the 45 lotteries for projects included in both our application sample and our survey sample, as well as for numerous other sampling restrictions. Empirically, a given household’s lottery participation is highly unpredictable.
Table 3: Identification tests for survey response: Regression estimates

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dependent variable: I{survey respondent}</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Std. parking ratio</td>
<td>-0.0001</td>
<td>-0.005</td>
<td>-0.003</td>
<td>0.003</td>
</tr>
<tr>
<td></td>
<td>(0.010)</td>
<td>(0.012)</td>
<td>(0.012)</td>
<td>(0.011)</td>
</tr>
<tr>
<td>Std. Transit Score</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>-0.011</td>
<td>-0.013</td>
<td>-0.004</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.017)</td>
<td>(0.023)</td>
<td>(0.021)</td>
<td></td>
</tr>
<tr>
<td>Std. Walk Score</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>-0.013</td>
<td>-0.021</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.013)</td>
<td>(0.013)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Std. Bike Score</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>0.022</td>
<td>0.026</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.015)</td>
<td>(0.014)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Average response rate</td>
<td>0.294</td>
<td>0.294</td>
<td>0.294</td>
<td>0.294</td>
</tr>
<tr>
<td>Controls</td>
<td>No</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
</tr>
<tr>
<td>Observations</td>
<td>2,654</td>
<td>2,654</td>
<td>2,654</td>
<td>2,654</td>
</tr>
</tbody>
</table>

*p < 0.05; **p < 0.01; ***p < 0.001  Notes: Standard errors clustered by residential project. Controls are household income, household size, survey recipient gender, and survey recipient race. Each observation is a household to whom we (e)mailed a survey. Each column presents estimates from a separate regression model.
Figure 4: Survey responses for car ownership by residential parking ratio

On site parking spaces per residential unit

Car ownership rate (percent)

Source: Authors’ analysis of MOHCD data and authors’ survey.
Notes: Each bar contains a set of BMR projects binned by the on site residential parking space ratio, with bars’ heights corresponding to the share of included survey respondents that own any automobiles.
walking, and bicycling, along with survey respondent-level controls. Transit accessibility emerges as a somewhat smaller influence on car ownership, and is insignificant in Column (4), likely because of the strong collinearity noted above between transit, walking and bicycling accessibility. On the whole, car ownership appears to be strongly influenced by features of the local built environment.\(^{13}\)

Table 4: Survey responses for car ownership: Regression estimates

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dependent variable:</td>
<td>I{own any cars}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Std. parking ratio</td>
<td>0.143***</td>
<td>0.121***</td>
<td>0.124***</td>
<td>0.123***</td>
</tr>
<tr>
<td></td>
<td>(0.023)</td>
<td>(0.026)</td>
<td>(0.027)</td>
<td>(0.028)</td>
</tr>
<tr>
<td>Std. Transit Score</td>
<td>-0.048***</td>
<td>-0.037**</td>
<td>-0.019</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.012)</td>
<td>(0.014)</td>
<td>(0.023)</td>
<td></td>
</tr>
<tr>
<td>Std. Walk Score</td>
<td></td>
<td></td>
<td>-0.037</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(0.030)</td>
<td></td>
</tr>
<tr>
<td>Std. Bike Score</td>
<td></td>
<td></td>
<td>0.021</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(0.034)</td>
<td></td>
</tr>
<tr>
<td>Average car ownership</td>
<td>0.668</td>
<td>0.668</td>
<td>0.668</td>
<td>0.668</td>
</tr>
<tr>
<td>Controls</td>
<td>No</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Observations</td>
<td>758</td>
<td>758</td>
<td>758</td>
<td>758</td>
</tr>
</tbody>
</table>

*p<0.05; **p<0.01; ***p<0.001  Notes: Standard errors clustered by residential project. Controls are household income, household size, survey recipient gender, and survey recipient race. Each column presents estimates from a separate regression model.

In addition to impacting car ownership, parking ratios and transportation accessibility also affect household transportation mode decisions. Figure 5 shows the raw correlations between project- and neighborhood-level transportation availability characteristics (rows) and surveyed households’ travel behavior (columns). As expected, the frequency of driving (bottom row) increases with the building’s parking ratio and decreases with neighborhood transit, walking, and

\(^{13}\)Household decisions pertaining to car ownership are also likely to be affected by the price of residential parking, which is strongly related to supply. In our applicant sample, 76 percent of successful lottery applicants were offered a parking space, but only 28 percent of them accepted a space. Low acceptance rates are unsurprising given the cost of parking ranges from $100 to $350 per month for the rental units in our applicant sample, and from $33,000 to $138,124 as a one-time payment for the ownership units.
cycling accessibility. The frequency of bicycling, walking, and transit use (the top three rows) show the opposite relationship to that for driving. Across the board, these correlations strongly support the conclusion that households choose between driving and other modes of travel based on the quality and availability of local modes of transportation.

Figure 5: Survey responses for household transportation utilization: Correlation matrix

<table>
<thead>
<tr>
<th></th>
<th>Parking ratio</th>
<th>AllTransit Score</th>
<th>Walk Score</th>
<th>Bike Score</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bike freq.</td>
<td>−0.05</td>
<td>0.06</td>
<td>0.09</td>
<td>0.10</td>
</tr>
<tr>
<td>Walk freq.</td>
<td>−0.14</td>
<td>0.20</td>
<td>0.20</td>
<td>0.15</td>
</tr>
<tr>
<td>Transit freq.</td>
<td>−0.23</td>
<td>0.16</td>
<td>0.18</td>
<td>0.13</td>
</tr>
<tr>
<td>Drive freq.</td>
<td>0.20</td>
<td>−0.19</td>
<td>−0.21</td>
<td>−0.14</td>
</tr>
</tbody>
</table>

Source: Authors’ analysis of MOHCD data and authors’ survey.
Notes: Surveyed transportation utilization frequencies are measured in increasing order using a discrete four-point scale of: “less often,” “2-4 times a month,” “2-3 times a week,” or “daily.”

The raw correlations provide compelling evidence that transportation choices depend on features of the local built environment. To more formally estimate the importance of these transportation availability measures in shaping households’ choices, we present multivariate regression analysis in Table 5. The dependent variables are a respondent’s self-reported frequency of travel by single-occupant vehicle, public transportation, walking, and bicycle, respectively. The survey
asked how often the respondent travels by each mode, on a 1 to 4 ordinal scale where a value of 4 is “daily,” a value of 3 is “2-3 times a week,” a value of 2 is “2-4 times a month,” and a value of 1 is “less often.” As expected, increasing accessibility by transit, walking, or bicycling increases the frequency of use of the corresponding mode, even after controlling for respondents’ income and demographics, as well as for the building’s parking ratio. Nearly all of the estimates are statistically significant at the five percent level, and most magnitudes are nontrivial. A one standard deviation increase in the building’s Walk Score, for instance, causes about a 19 percentage point increase in the likelihood that a household’s walking behavior falls into a more frequent bin.

Table 5: Survey responses for transportation utilization: Regression estimates

<table>
<thead>
<tr>
<th></th>
<th>Drive freq. (1)</th>
<th>Transit freq. (2)</th>
<th>Walk freq. (3)</th>
<th>Bike freq. (4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Std. parking ratio</td>
<td>0.218***</td>
<td>−0.269***</td>
<td>−0.138*</td>
<td>−0.023</td>
</tr>
<tr>
<td></td>
<td>(0.062)</td>
<td>(0.051)</td>
<td>(0.062)</td>
<td>(0.033)</td>
</tr>
<tr>
<td>Std. Transit Score</td>
<td>−0.141**</td>
<td>0.092*</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.045)</td>
<td>(0.044)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Std. Walk Score</td>
<td></td>
<td></td>
<td>0.185**</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(0.057)</td>
<td></td>
</tr>
<tr>
<td>Std. Bike Score</td>
<td></td>
<td></td>
<td></td>
<td>0.084***</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>(0.022)</td>
</tr>
</tbody>
</table>

Dep. var. average      | 2.195           | 2.627             | 2.875           | 1.351          |
Controls                | Yes             | Yes               | Yes             | Yes            |
Observations            | 766             | 766               | 766             | 766            |

Notes: Standard errors clustered by residential project. Controls are household income, household size, survey recipient gender, and survey recipient race. Each dependent variable frequency measure is treated as a continuous outcome formed from a four-point scale where a value of 4 is daily, a value of 3 is 2-3 times a week, a value of 2 is 2-4 times a month, and a value of 1 is less often. Each column presents estimates from a separate regression model.

In the case of public transit use, a building’s parking ratio also has an effect—and one that is approximately three times as large as that of transit accessibility. More on-site parking reduces transit use while increasing the frequency of driving by a similar amount. The parking ratio also
has a smaller but still statistically significant negative impact on the frequency of walking. Our estimates show no detectable impact of on-site parking on bicycling, although bicycling frequency is low for this population—83 percent of respondents report bicycling “less often” than 2-4 times a month.

The analysis above refers to all trips made by survey respondents. Similar patterns are shown in Table 6 for commute trips to work or school. Increased residential parking leads to a higher probability of commuting by private car (driving alone or carpooling) and a lower probability of commuting by transit. Greater transit accessibility has the opposite effects, although these results are not always as statistically significant.

Table 6: Survey responses for commuting transportation mode: Regression estimates

<table>
<thead>
<tr>
<th></th>
<th>Dependent variable: primarily commute by</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Private car (1)</td>
</tr>
<tr>
<td>Std. parking ratio</td>
<td>0.063* (0.029)</td>
</tr>
<tr>
<td>Std. Transit Score</td>
<td>-0.077*** (0.013)</td>
</tr>
<tr>
<td>Std. Walk Score</td>
<td>0.044*** (0.011)</td>
</tr>
<tr>
<td>Std. Bike Score</td>
<td>0.027*** (0.007)</td>
</tr>
<tr>
<td>Dep. var. average</td>
<td>0.384</td>
</tr>
<tr>
<td>Controls</td>
<td>Yes</td>
</tr>
<tr>
<td>Observations</td>
<td>544</td>
</tr>
</tbody>
</table>

*p<0.05; **p<0.01; ***p<0.001   Notes: Standard errors clustered by residential project. Controls are household income, household size, survey recipient gender, and survey recipient race. Each of these regressions includes only respondents who are either employed or a student. Each column presents estimates from a separate regression model.

The impact of parking and transportation accessibility on commute mode choice appear to be more muted than for non-work trips. This might be because commute trips are relatively more constrained, for example by workplace parking options or transit proximity, whereas non-
commuting trips entail more choice of potential destinations for (say) shopping or recreation. Another constraint relates to long distances that may preclude walking or bicycling. For this reason, Column (5) of Table 6 restricts the sample to only commutes made either by transit or private car. These estimates show even more clearly that commuters substitute between driving and transit based on the building’s parking ratio and transit accessibility. Also note that the outcome measures for all trips and commute trips are not directly comparable. Table 5 considers ordinal frequencies of use of all modes for all purposes, whereas Table 6 uses binary outcomes for respondents’ primary mode of commuting.

4.3 Survey analysis for employment

Finally, we evaluate employment outcomes and focus on two key transportation factors that the literature suggests may affect labor market opportunities, particularly for low-income workers. Access to public transportation and to private vehicles have both been found to improve employment outcomes, although the empirical evidence is rather mixed (Sanchez, 1999; Blumenberg and Ong, 2001; Sanchez et al., 2004; Grengs, 2010; Blumenberg and Pierce, 2014). For our surveyed households who are essentially randomly-assigned to a residential location, Table 7 suggests that neither transit accessibility nor parking ratios have any impact on the probability of a respondent being employed full-time (Column (1)). There is a similar null relationship with other labor market outcomes in Columns (2) to (4). One possibility is that these estimates are only indicative of the strong economy and minimal unemployment in San Francisco at the time of our survey in 2019. An alternative explanation is that marginal changes in car ownership and transit access are not very relevant for employment prospects after residential self-selection is fully accounted for via as-good-as randomization. For example, residing in a low-accessibility neighborhood might be correlated more generally with being unemployed because of some third factor such as discrimination in both housing and employment markets.

Another measure of labor market outcomes is provided by employment turnover. Greater accessibility may enable workers to change jobs, and in doing so increase wages, reduce commute time, or otherwise increase employment satisfaction. Column (1) of Table 8 shows the impact of parking ratios and transit accessibility on the share of employed/student respondents who have been at their current job or school for less than two years; Column (2) does the same for a shorter one-year period. The remaining columns measure the impacts on commute distance and time directly.

Greater availability of on-site parking has no detectable impact on any of these labor market outcomes, although greater transit accessibility appears to moderately promote employment
Table 7: Survey responses for employment status: Regression estimates

<table>
<thead>
<tr>
<th>Dependent variable</th>
<th>Work full-time</th>
<th>Work part-time</th>
<th>Looking for work</th>
<th>Student</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
</tr>
<tr>
<td>Std. parking ratio</td>
<td>0.018 (0.018)</td>
<td>−0.004 (0.014)</td>
<td>−0.003 (0.007)</td>
<td>−0.008 (0.010)</td>
</tr>
<tr>
<td>Std. Transit Score</td>
<td>−0.010 (0.009)</td>
<td>0.004 (0.006)</td>
<td>0.003 (0.002)</td>
<td>0.001 (0.007)</td>
</tr>
</tbody>
</table>

| Dep. var. average | 0.865 | 0.095 | 0.023 | 0.026 |
| Controls          | Yes   | Yes   | Yes   | Yes   |
| Observations      | 660   | 660   | 660   | 660   |

Notes: Standard errors clustered by residential project. Controls are household income, household size, survey recipient gender, and survey recipient race. Each of these regressions includes only respondents who are either employed, looking for work, or a student. Each column presents estimates from a separate regression model.

Table 8: Survey responses for employment duration and commutes: Regression estimates

<table>
<thead>
<tr>
<th>At current work/school</th>
<th>Current commute via driving</th>
</tr>
</thead>
<tbody>
<tr>
<td>&lt; 2 years</td>
<td>Distance (m)</td>
</tr>
<tr>
<td>(1)</td>
<td>(2)</td>
</tr>
<tr>
<td>Std. parking ratio</td>
<td>−0.011 (0.018)</td>
</tr>
<tr>
<td>Std. Transit Score</td>
<td>0.023* (0.009)</td>
</tr>
</tbody>
</table>

| Dep. var. average | 0.222 | 0.110 | 6,748 | 13.05 | 0.11 |
| Controls          | Yes   | Yes   | Yes   | Yes   | Yes   |
| Observations      | 644   | 644   | 634   | 634   | 634   |

Notes: Standard errors clustered by residential project. Controls are household income, household size, survey recipient gender, and survey recipient race. Each of these regressions includes only respondents who are either employed or a student. The current commute distance is the estimated driving distance in meters and the current commute driving time is estimated as of 8:00 AM on a weekday using Google Maps. Each column presents estimates from a separate regression model.
turnover and shorter current commute times.\textsuperscript{14} A one standard deviation improvement in Transit Score increases the share of respondents occupied at their current job or school for less than two years by 2.3 percentage points—a meaningful effect given that only 22 percent of respondents have been at their workplace for such a short period. The results are almost identical when limiting the sample to respondents in full-time work. Similarly, we find that a one standard deviation increase in Transit Score reduces commuting times by about ten percent—and reduces the likelihood of having a long (> 25 minutes) commute by about 30 percent. In other words, greater transit accessibility increases the likelihood of garnering new employment and reduces commute times.\textsuperscript{15}

5 Conclusions

In this report, we use San Francisco’s residential housing lotteries to study how local parking and transportation accessibility affect household behavior. Because the odds of winning any specific lottery are low and there are no monetary costs of entry, households are understandably quite unselective about which lotteries they enter. As we demonstrate, those who are fortunate to win any lottery are thus as-good-as-randomly assigned into living in particular residential locations. In contrast to nearly all research on the implications of transportation accessibility for travel and employment outcomes, which is susceptible to selection bias from residential sorting, our findings have a causal interpretation.

Our evidence is limited to households that are eligible for affordable housing programs. In San Francisco, however, these programs target a wide income range: households earning up to 120 percent of median income ($118,200 for a two-person household in 2019) are often eligible for these housing lotteries.

We find that neighborhood-level accessibility has statistically significant and quantitatively meaningful impacts on household decisions about car ownership and travel. Greater transit accessibility reduces the propensity to own and drive a car, while increasing the propensity to ride transit. Greater walk and bicycle accessibility also increase the propensity to use those modes. These findings are not surprising, but confirm that the land use-transportation relationships commonly shown in this literature are not simply a product of self-selection and other biases. Even within San Francisco, transit accessibility substantially affects car ownership and travel behavior—increasing

\textsuperscript{14}Estimates show no relationship between parking or transit accessibility and former workplace/school commutes or with the change from former to current workplace/school commutes, in time or distance.

\textsuperscript{15}Responses for our measure of social capital (survey question 7 in Figure 2) are largely unrelated to transportation availability, other than a small and not very robust positive association with the building’s Walk Score.
transit accessibility from the level of an outer suburban neighborhood (the 5th percentile) to the
citywide median would increase the share of those commuting by transit rather than by car by
6.5 percentage points. San Francisco is a more walkable, bikeable, and transit-accessible city
compared to most locations, suggesting that even more substantial household responses to in-
creased bus scheduling, for example, might be expected in places where transit service is minimal
at present.

We also document a more novel relationship between the residential parking provided in a
building and transportation outcomes. Given that households who wish to own a car likely have
numerous external parking options—to park on-street, park in a public garage, or rent a space in
a nearby building—one might surmise that neighborhood-level rather than building-level parking
supply would most affect transportation outcomes. However, we show that a building’s parking
ratio not only influences car ownership, vehicle travel, and transit use, but has a stronger effect
than transit accessibility. Buildings with at least one parking space per unit (as required by zoning
codes in most U.S. cities, and in San Francisco until circa 2010) have more than twice the car own-
ership rate of buildings that have no parking. If parking is provided on-site for free or at a reduced
price (typically, $100 per month), then households appear to take advantage of this amenity. In
contrast, households without access to on-site parking are more likely to forgo car ownership al-
together.

One natural concern about reducing required parking ratios is that this might limit employ-
ment opportunities, particularly for lower-income households such as those we study in this re-
port. Given that many jobs are inaccessible by public transit, access to a car can theoretically
improve employment outcomes and labor market turnover. However, we find no evidence that
this tradeoff exists.

Transit accessibility evolves over decades and a concerted effort to improve local infrastruc-
ture requires large amounts of public funding. Parking ratios, in contrast, require only regulatory
changes to zoning codes: removing minimum requirements from zoning codes and possibly re-
placing them with maxima instead. Such zoning reforms could also yield other benefits including
reducing housing costs and increasing land available for new housing and commercial develop-
ment, as well as reducing motor vehicle trips and associated harms. Our findings suggest that the
potential for private automobile trip reductions is large and does not depend on car-free house-
holds relocating to car-free buildings. Moreover, reducing space dedicated to parking appears to
come without employment downsides. Where streets are relatively walkable and transit service
is frequent, parking emerges as the key factor shaping household travel behavior—and parking is
a factor that is highly amenable to low-cost policy reforms that can rapidly provide benefits.
References


