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4 **Location Choice vis-à-vis Transportation: The Case of Apartment Dwellers**
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33
34 **Abstract**

35 An understanding of residential location choice is fundamental to behavioral models of land use
36 and, ultimately, travel demand. Detailed data and predictive models are lacking. The paper
37 examines the choices of apartment dwellers and explores their reasons for moving, priorities
38 when choosing a residential location, and tradeoffs involved. In addition to summary statistics of
39 the data, linear regressions, binary logit, and ordered probit models were utilized to investigate
40 variations in rent and apartment size, stated preferences of housing, location, transportation, and
41 access. Binary logit and ordered probit models reveal similar results concerning people's
42 preferences for accessibility. For instance, families and other multi-person households tend to
43 place less value on commute times and freeway access and choose apartment improvements over
44 travel savings. Interestingly, women are more likely to state that they place a higher importance
45 on commute time and freeway access; but, when asked to choose between travel times and
46 apartment size, they are more likely to choose the larger apartment. Other models suggest that
47 being within walking distance of a commercial center increases average rent by \$24 per month.
48 Increases in distances to the central business district (CBD) and mean neighborhood commute
49 times reflect lower monthly rents, about \$20 per mile from the CBD and \$24 per added minute of
50 commute (one-way). Apartments in the urban area tend to be, on average, 75 square feet smaller,
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3 ceteris paribus (including population density, which has an added effect). These results and many
4 others provide several valuable insights regarding the location choice of those residing in
5 apartments.
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7 **Keywords:** Location choice, logit models, apartment choice, accessibility
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10 **1 Introduction**

11 The past 40 years have seen significant urban shifts in land use and travel behaviors. Rising
12 income and vehicle ownership have made it possible for many families to purchase apartments in
13 suburban areas and travel longer distances, resulting in minimal transit use and decentralization
14 of metropolitan areas. Such shifts make integrated models of land use and transportation very
15 relevant for prediction of future travel patterns. Residential location choice models can inform
16 such models.
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18 This paper focuses on apartment dwellers in order to obtain a clearer picture of the underlying
19 factors for choosing for their residential choices, vis-à-vis many factors. According to the Census
20 of Population Survey (CPS), renters comprised 62.7% of movers during 2002 and 2003.
21 (Schachter 2004) Though they represent the majority of movers, they only represent 33.8% of
22 U.S. households. And they are a demographic group that has not previously been studied in
23 much detail. This research developed a survey instrument that asked randomly selected
24 apartment residents in Austin, Texas about their reasons for choosing to live in an apartment and
25 for moving, the importance they place on certain housing and location attributes, their travel
26 patterns, their opinions and values, and basic demographic information. The remainder of the
27 paper positions the study within the context of prior work, describes the methodologies
28 employed, and discusses summary statistics of data collected as well as empirical results of
29 linear regression and discrete choice models. Key results and extensions are discussed in the
30 conclusions, providing a platform for future research.
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32 **2 Literature Review**

33 The standard framework for residential location choice models hypothesizes a sequence of
34 decisions that begins with a decision to move and ends with a chosen home and location. (Grigg
35 1982, Weisbrod et al. 1980, Guiliano 1988, Ben-Akiva and Bowman 1998) Studies have
36 examined various aspects of residential location choice, such as residential mobility (Speare et
37 al. 1975), market search (Clark 1982), dwelling type (Boehm 1982, Tu and Goldfinch 1996, Cho
38 1997), and location choice (Gabriel and Rosenthal 1989, Wadell 1996). These models seek to
39 identify the determinants of household mobility as well as choice of apartment and location.
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42 Although there are many residential location choice models, most do not identify reasons to
43 move. The US Bureau of Census recognized this gap and recently published a couple Current
44 Population Reports titled “Why People Move” (2001) and “Geographic Mobility” (2004),
45 containing cross-tabulations and raw distributions. These studies included a high number of
46 reason-to-move responses in “other” categories, suggesting that there are some unexpected yet
47 important reasons for moving. The studies did not quantify correlations between multiple
48 demographic factors and response nor did they identify the type of housing structure or tenure
49 choice.
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4 A few residential location choice studies did include reasons for moving. Murie's 1974 study in
5 England explored the reasons for household move and related them to tenure, housing structure,
6 and several demographic factors, but the data is out-dated and the housing and tenure options are
7 very different from the dominant types of current housing. Filion et al. (1999) extensively
8 investigated the determinants of residential location choice within Kitchener, Canada. They
9 reported households' reasons for moving but did not relate these to housing structure or
10 demographics. To the authors' knowledge, no recent study exists that isolates apartment dwellers
11 and explores their reasons for moving.
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13 Another important aspect of residential location choice involves the housing search process,
14 particularly the relative importance of various attributes. Filion et al. (1999) presented some raw
15 statistics. However, their study did not explore explanatory variables that underlie the varying
16 importance of such attributes. The 2004 American Community Survey (ACS) also examined
17 household priorities when deciding where to live. Although Belden et al. (2004) linked gender
18 and race in the ACS, they presented little analysis and did not relate such priorities to dwelling
19 type.
20

21 The third aspect of residential location presented in this paper concerns the tradeoffs that
22 households make when choosing an apartment. A household's choice to move and where to
23 move is a complex and costly decision. "When people buy or rent housing, they are obtaining a
24 bundle of goods that includes interior living space; housing services such as schools and parks;
25 and externalities like neighborhood image, noise, and smog." (NCHRP Report 423A 1999, p.96)
26 For virtually every household, a residence cannot be found in which all of these housing and
27 location attributes are optimized; and size, cost, accessibility, or other features may be
28 compromised. Weisbrod et al. (1980) examined the tradeoffs between transportation and other
29 factors for recent movers in Minnesota. Although they did calibrate a tenure choice model, they
30 did not quantify tradeoffs for apartment dwellers nor link demographic characteristics to these.
31 Belden et al. (2004) explored tradeoffs between commute time and lot size while linking gender
32 and race. However, they discussed only raw statistics.
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34 Overall, the research presented in this paper is unique in that it focuses on apartment dwellers
35 and addresses their reasons to move, their valuation of various factors while searching for a new
36 apartment, and the tradeoffs associated in apartment choice and location.
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38 **3 Methodology**

39 Survey design and data collection were undertaken by graduate students at the University of
40 Texas at Austin during the spring semester of 2005 as part of a collective effort between
41 researchers and students in a graduate course. The survey was designed as a self-completion
42 survey and was intended for door-to-door as well as Internet distribution. Several revisions and a
43 pilot test were executed in order to develop a comprehensive survey, which can be found in Bina
44 (2005).
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46 **3.1 Sampling**

47 The sampling frame for the survey was all apartment dwellers within the Austin area¹. The 2000
48 Census estimates 138,757 renter-occupied multi-unit attached housing. A list of 558 apartment
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3 complexes (representing 115,344 apartments) was obtained from Austin Investor Interests and
4 the University of Texas at Austin Division of Housing and Food Service datasheet. Thus, the
5 sampling list obtained seems to be fairly comprehensive (containing 83% of all such units) and is
6 hopefully, unbiased.
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9 Due to resource limitations, a stratified cluster sampling approach was used to select apartment
10 complexes. The stratification recognized four regions of roughly equal populations (200,000
11 persons). It also recognized complex size since complexes of similar size may be alike in terms
12 of amenities, which can be important to renters and yet hard to quantify. Thus, sampled
13 complexes were chosen randomly with equal numbers of “small” (80 or fewer rentable units),
14 “medium” (81 to 250 rentable units), and “large” (greater than 250 rentable units) complexes.
15 (The average complex size is roughly 200 apartments.) Six complexes (two of each “size”) were
16 selected for each of the four regions. However, since data collectors were required to receive
17 only 40 completed surveys and some fulfilled the quota before sampling every complex, only 17
18 complexes were actually surveyed. Supplementary data was obtained to describe each
19 observation’s location. Capital Area Metropolitan Planning Organization (CAMPO) data
20 provided information on zonal areas, population, number of households, and employment at the
21 Traffic Serial Zone (TSZ) level; and Census tract information on housing characteristics was
22 matched to the TSZ.
23

24 **3.2 Survey method**

25 After running a pre-test of the survey instrument using 10 demographically diverse apartment
26 dwellers, the survey was distributed “door-to-door” on Saturdays and Sundays during late
27 February and early March of 2005. The survey was delivered directly to the first adult answering
28 the door and collected from respondents around 30 minutes later. The reasons for choosing this
29 survey method are several: This method permitted faster distribution and response times, as well
30 as higher response rates (Richardson et al 1995). It also permitted better data quality by allowing
31 respondents to get their questions answered directly. Candy bars and maps were offered as
32 incentives, and cards advertising the website URL were posted at unopened doors.
33

34 **3.3 Response rates**

35 A total of around 1600 apartments were visited; out of these, 28% answered the door. Only 450
36 doors were opened, perhaps because no one was home, lived, or wished to answer the door at the
37 others. This is largely a quality neutral loss, though certain travel, location choice or other
38 relevant characteristics may be associated with those living in the non-response apartments. The
39 surveys were conducted on weekend days only, when most people, regardless of employment
40 type, may be assumed to have the same chance of being at home.
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42 Of the 450 who answered the door, 260 chose to return a survey, suggesting a response rate of
43 58%. However, only 240 of those surveys were fully completed. So the real response rate was in
44 fact 53%. Generally, women were more likely to answer the doors than men, and younger
45 persons were more likely to answer the door than older persons. Among the women who
46 answered their doors, more than half agreed to fill out the survey, while slightly less than half of
47 the men agreed. Elderly persons appeared much more reluctant to take the survey than younger
48 people. Also of some interest is the fact that both men and women were more responsive when a
49 person of the opposite gender was asking, even in cases where there were two students of
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3 different gender interviewing at the same time (with one standing in the background). In such
4 cases, the female interviewer tended to achieve higher response rates, confirming previous
5 response rate studies.
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7 8 **3.4 Weights**

9 Several of the 240 “completed” surveys required some data imputation (as discussed below).
10 Weights to correct for age, gender, and household income were created using most recent 5%
11 Public-Use Microdata Sample (PUMS) for Austin metro area renters in apartment buildings (not
12 including those in institutionalized group housing units or those under the age of 18). The sample
13 weights were created for 18 groups of people, as characterized by 3 age groups (18-35, 36-55,
14 56+ years of age), 3 household income groups (\$0-\$24,999, \$25,000-\$49,999, \$50,000+), and
15 gender.
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17 **3.5 Imputed data values**

18 Where feasible, missing data was imputed. For example, rents were determined by comparing
19 apartment units with others obtained from the same apartment complex. In many cases these
20 were virtually identical. When rent values varied across a complex, comparisons based on rent
21 per square foot as a function of bedrooms and bathrooms provided a clear indication of the
22 appropriate rent category.
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24 Square footage was imputed similarly, recognizing the number of rooms and rent levels within
25 each apartment complex. However, since the variation of square footage within each apartment
26 is much greater than rent variations (possibly due to the respondents’ ignorance of exact square
27 footage, as compared to rent), some values could not be imputed with sufficient certainty and
28 remained missing.
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30 Missing values for respondent age were imputed using ordinary least squares (OLS) regression
31 techniques. A two-sample t-test suggested that age values were missing at random across
32 observations. Stochastic regression imputation was used².
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34 As with many surveys, many household income responses were missing. Since this variable was
35 reported categorically (i.e., as “grouped data”), a multi-threshold variation of the tobit model was
36 used in LimDep software in order to provide an underlying continuous model for income
37 prediction. These continuous values were then used for missing values, while category mid-
38 points were used for all reporting households.
39

40 **4 Data Analysis and Results**

41 The following discussion presents sample characteristics and results of behavioral regression
42 models. Table 1 provides several summary statistics that characterize apartment dwellers in the
43 sample.
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45 Many practitioners and researchers are interested in why a household chooses a particular
46 dwelling type. The survey asked the respondents to indicate their main reason for choosing to
47 live in an apartment. 44% indicated affordability, 18% needed a short-term residence, 15%
48 appreciated the size, relative to their needs, 13% wanted low maintenance, and 9.5% chose
49 “other” as a response. Based on these responses, one might hypothesize that lower income and
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3 smaller households tend to live in apartments. 2000 Census PUMS data for the Austin metro
4 area confirms this hypothesis, indicating that the average household income of those living in
5 apartments is \$35,996 – or less than half that of non-apartment dwellers (\$74,163). Moreover,
6 the average household size for those residing in apartments is 2.08 persons, whereas an average
7 of 2.63 persons live in other types of dwelling units.
8

9 10 **4.1 Reasons for Moving**

11 Simply knowing why people move can be very helpful in developing residential choice models.
12 The survey asked respondents to indicate their primary reason for moving to their current
13 apartment. Table 2 compares these results to those of the 2003 U.S. CPS, which sampled over
14 40,000 recently relocated households across the U.S.
15

16 The comparisons suggest that Austin’s apartment dwellers differ from recent U.S. movers in
17 several ways. The greatest difference between the two is the high percentage of apartment
18 dwellers surveyed that moved for an easier commute. This may be attributed to Austin’s heavy
19 congestion and limited freeway corridors. The next greatest difference relates to those moving
20 for a new job/job transfer: 4.77% more apartment dwellers stated this as their primary reason for
21 moving. A new job or job transfer often signals a long-distance move; and the Census results
22 support this by indicating that the most common single reason for an intercounty or international
23 move is a new job or job transfer. (Schachter 2004) Long-distance movers may be more inclined
24 to choose an apartment, in order to become more familiar with the area before buying a home. A
25 third difference is the higher percentage of apartment dwellers seeking less expensive housing,
26 which is intuitive since apartments are generally a less expensive housing option. Finally, a
27 higher percentage of apartment dwellers moved to begin college studies, which also is intuitive,
28 since many college students rent apartments and Austin has a relatively high population of
29 college students (13.7% vs. 8.32% in the US).
30

31 **4.2 Priorities during Housing Search**

32 Once a household has chosen to move, the process of searching for a new apartment/location
33 begins. During this search, a household has priorities for key features. So respondents were
34 asked to rank the importance of several housing and location attributes. Table 3 lists these
35 attributes, along with the “mean” ranks for the corrected (population weighted) sample.
36

37 Predictably, price is the important attribute to apartment dwellers. Of course, price is a key
38 criterion in virtually any choice, for most people. Moreover, lower income households tend to
39 rent (as discussed earlier), and therefore may be more concerned with this attribute. Commute
40 time is the next most important attribute, which, as explained earlier, may be credited to Austin’s
41 traffic congestion. Commute time is just one of several access attributes that were included in the
42 survey. By summing the weights of all variables, access attributes carry less importance than
43 non-access attributes (40% vs. 60%).
44

45 Surprisingly, the quality of and distance to local public schools attributes were rated least
46 important. Perhaps this is because apartment dwelling households tend to contain fewer children.
47 The 2000 Census suggests that 20.4% of U.S. households living in an apartment have children,
48 as compared to 30.3% among non-apartment households. Ordered probit models were created to
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3 analyze the underlying factors that influence these scores. And the presence of children was a
4 statistically significant variable in some cases.
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6 **5 Model results**

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8 Weighted least squares (WLS), binary logit, or ordered probit regression models were used to
9 analyze response to the various types of survey questions posed. The results are as follows:

10 **5.1 Linear regression analyses of rent and square footage**

11 Linear regression models (Table 4), weighted by population correction factors, were used to
12 examine how rent and square footage relate to various demographic and location variables. This
13 is valuable information in determining where to build and zone for multifamily apartment
14 complexes (as well as how to price such units). The results also provide a sense of the tradeoffs
15 that households make in terms of cost (rent) and benefits (e.g., interior square footage). As
16 shown in Table 4, all variables that were expected to have an impact were included in the initial
17 specifications. The final model specifications emerged from a systematic procedure of
18 eliminating statistically insignificant variables, combined with intuitive considerations. Final
19 adjusted R^2 values exceeded 0.5, suggesting a reasonable fit – but also the fact that many other
20 variables are at play here.
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22 **5.1.1 Rent model**

23 The average rent in the dataset was \$693 per month. Each added bedroom's estimated value is
24 \$119, and each bathroom \$109. While an added bedroom may be more useful to many
25 households and offer more space than a bathroom, bathrooms are expensive to build and service.
26 Having a commercial center within walking distance adds around \$24 per month in rent. And
27 brand new apartments are expected to command \$44 more per month.
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29 Non-Caucasian households tend to pay \$52 less per month, while those with children tend to pay
30 around \$47 less per child. Those with higher levels of education tend to pay more (e.g.,
31 \$110/month by those with a Master's degree). Such attributes may be proxying for location
32 effects not captured by other model variables. These other variables include proximity to the
33 CBD, which is valued quite favorably: Every mile less in travel distance to the CBD contributes
34 an average of \$20 in monthly rent. A similar trend is visible in the mean-travel-time-to-work
35 variable: For every minute less of commute time, rents rise by \$24/month.
36

37 Rents also tend to rise with population density, ceteris paribus: Another 3,000 people per square
38 mile (or 4.7 persons per acre) is associated with rents that are \$55 per month higher. However,
39 increased transit stop density counters this effect: Another 50 bus stops per square mile averages
40 \$67 less in monthly rent. This may due to the fact that the use of bus transportation is more
41 widespread among lower income households. It also may relate to a greater presence of
42 commercially used, busy streets, where bus stops are common, but noise, congestion, and other
43 issues limit desirability for residential use. Many of these same features are at play in apartment
44 size estimation, as discussed next.
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46 **5.1.2 Square footage**

47 The WLS model of apartment size suggests that another bedroom adds around 152 square feet,
48 and an extra bathroom 179 square feet. Since bedrooms tend to be quite a bit larger than
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bathrooms, this result is most likely an indication that the overall size of an apartment is influenced by the number of bathrooms. In other words, the model specifications does not suggest that bathrooms have an average size of 179 square feet; but, rather, having more than one bedroom may be an indicator of a “luxury” apartment, offering more space throughout the unit.

Households with children appear to use less space, dropping about 22 square feet per child, which is not an intuitive result. One would expect families with children to require more space. However, it could be an indication that families with many children have tighter budgets and thus they are forced to select smaller apartments, everything else constant. This is consistent with the results of WLS models of rent, in section 5.1.1, which suggests that families with children pay less in rent than childless households. Since children add more expenses to the family, such households cannot necessarily afford as expensive (and large) an apartment as households without children. This conclusion is further reinforced when one looks at higher-income households. They tend, *ceteris paribus* to choose more spacious apartments (0.77 square feet more per \$1,000 in annual income). Respondents with master’s degrees or higher levels of education tend to live in apartments that average an additional 94 square feet.

As expected, smaller apartments are found in Austin’s “urban areas” (70 square feet less than in non-urban areas, as defined by CAMPO). Higher population densities are associated with smaller apartments, as expected: Another 3,000 persons per square mile is associated with 130 less square feet. Interestingly, after controlling for these two types of variables, size is estimated to fall with distance from the CBD (at a rate of 37 square feet per mile). This may indicate that those willing to pay to live more centrally also want larger units. Access and size both come at a price, however, as discussed earlier.

5.2 Logit results for binary choice experiments

The six stated preference questions were developed in order to appreciate which apartment respondents prefer. All six scenarios presented a choice between an improved apartment or neighborhood feature and a transportation improvement. The scenarios and their weighted choice percentages are as follows:

- Scenario 1: 200 extra SF (47%) vs. freeway proximity reducing commute time by half (53%).
- Scenario 2: An apartment with friend or relatives nearby (55%) vs. an apartment near a light rail station that can take the respondent to work or school (45%).
- Scenario 3: A suburban apartment with plenty of parking (66%) vs. a downtown apartment with one parking space (and additional parking spaces costing \$60 per month) (34%).
- Scenario 4: An apartment close to a shopping center (41%) vs. a larger kitchen/living room (59%).
- Scenario 5: An apartment close to a bus stop (46%) vs. one offering a park view (54%).
- Scenario 6: A brand new apartment and complex (77%) vs. an older apartment that is 5 miles closer to a shopping center (23%).

Table 5 shows the model results for the six scenarios. In every comparison, Apartment 2 is the base choice, meaning that the parameter estimates represent the additional utility of Apartment 1, as compared to Apartment 2. As before, elimination of statistically insignificant variables and

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3 intuitive considerations have been used to obtain the final specifications. A p-value of 0.20 was
4 generally accepted as the upper limit of statistical significance. However, the relatively small
5 sample sizes make it difficult to obtain statistical significance on all variables of interest.
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7 This section describes preferences by demographic groups, as revealed by the model results.
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9 **5.2.1 Household size and income**

10 Larger households and married couples tend to prefer larger apartments and more parking, as one
11 might expect, while single-person households are more likely to opt for a shorter commute time
12 and a downtown location. Larger households also tend to value apartment enhancements over
13 access improvements. Hence, they are more likely to choose better appliances and a newer
14 apartment than reduced shopping travel time (Scenario 6). Those with children are more likely to
15 opt for a nearby park (where their children can play, ostensibly) than transit access. Those with
16 many workers, however, are attracted by the light rail option. Higher-income households tend to
17 value a park view over bus stop proximity, and a newer complex over nearby shopping, perhaps
18 because travel costs (including parking) are of less importance to them.
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20 **5.2.2 Ethnicity and gender**

21 Ethnicity parameters emerge as statistically significant in four scenarios, but only when grouped
22 (as Caucasian and non-Caucasian). In general, the results suggest that non-Caucasian households
23 are more interested in shorter travel times (to shopping and workplaces) than in better apartment
24 features. This may indicate that these demographic groups depend more on public transportation
25 or other non-SOV modes, or it may be they are more time-constrained in their activities.
26

27 Women appear to prefer larger apartments, over reduced commute times, relative to male
28 respondents. That may be due to shorter commute times, on average, for women (their average
29 commute times are roughly the same: 21.82 minutes for men vs. 20.47 minutes for women).
30 Sermons and Koppelman's (2001) work suggests that women spend less time commuting due to
31 their greater participation in household activities.
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33 **5.2.3 Education and employment**

34 Education and employment status also affect respondent priorities. Scenarios 3 and 4 suggest that
35 more highly educated persons are more likely to choose reduced travel times (to shopping) and a
36 downtown location, possibly because they tend to work longer hours and/or have higher values
37 of time, ceteris paribus. Full-time workers also are more attracted to travel time savings, in their
38 commutes. And retired persons tend to be more impressed by shopping access (than by newer
39 apartments).
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41 **5.2.4 Apartment location**

42 In all six scenarios, supplementary data regarding current apartment location indicate that urban
43 area apartment dwellers are more likely to choose shorter commute times, better public
44 transportation facilities and proximity to shopping centers. Such households may be more
45 accustomed to using (and dependent on) public transit. The distance-to-CBD parameter suggests
46 that households located further from the CBD are more likely to opt for better public
47 transportation (bus and rail) options. This could be an indication that public transportation in the
48 suburbs does not meet the requirements of the citizens in those areas.
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5 **5.3 Ordered probit analysis of the importance of access**

6 Ordered probit models were used to explore priorities during the housing search process. Since
7 the variables of primary interest concern accessibility and its impact on location choice,
8 explanatory variables like commute time, distance/travel time to shopping, access to major
9 freeways, and access to public transportation were studied. Final model specifications are shown
10 in Table 6, and these provide some interesting results.

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12 Those who view commute time as more important tend to be female, non-Caucasian, highly
13 educated (master's degree or higher), and have no children. Among these, the presence of
14 children is the most practically significant, causing more than a one-point gain in terms of
15 importance (which is scored from 1 to 5). A graduate degree is almost as significant, in this
16 same sense.

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18 Those who view shopping access as more important tend to be older, Hispanic or Latino, having
19 fewer workers in the household, and living with family members (but not with a spouse and
20 children). Transit access is rated as more important by students, non-Caucasians, and those
21 with fewer vehicles, lower levels of education, and lower household income. Freeway access is
22 rated higher by females, Hispanics, Latinos and African -Americans, those of lower educational
23 attainment, and those without children at home. Those living with family and/or a significant
24 other are also more likely to rate freeway access highly.

25
26 These various attributes, and preferences, offer one a sense of the consumer market for different
27 locations.

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29 **5.4 Some potential applications of results**

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31 The results of these models tell a bigger story than simply who is more attracted to what and
32 what they are willing to pay. For example, the logit results suggest that if a developer and/or
33 community wishes to attract well-educated, high-earning full-time workers, it might best focus
34 on building nice apartments close to downtown, while improving access to public transportation.
35 In order to attract families with children, however, they should build large apartment complexes
36 in the suburbs with access to recreation facilities and shopping.

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38 Another possible goal of communities is greater ethnic and racial integration. Since non-
39 Caucasian respondents appear to value public transit access, improvements in bus and/or
40 additions of light rail service in neighborhoods dominated by Caucasian households may serve
41 such objectives. Rents should probably be kept moderate in enough units to ensure affordability
42 for a variety of household types.

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44 The model of rent arguably indicates substantial differences in willingness to pay. For example, a
45 white single person, with a graduate degree and an annual income of \$80,000 is estimated to pay
46 \$1216 per month for a single-bedroom, single-bathroom, new apartment, with a commercial
47 center nearby, one mile from the CBD, and a mean commute time of 10 minutes (and densities
48 of 3000 persons and 17 bus stops per square mile). In notable contrast, a non-Caucasian with
49 three children, a bachelor's degree or less, and an annual income of \$30,000 is willing to pay
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3 only \$874 per month for the very same apartment. Of course, an apartment of this type may not
4 be available at that price, suggesting that certain demographic groups will be priced out of this
5 market. Such distinctions support that notion that market forces can (and do) result in substantial
6 clustering of households, by income education transportation needs, and other factors.
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9 Finally, in order to deal with issues of congestion, transit-oriented designs that cater to a variety
10 of preferences may be of interest. By locating an apartment complex in the suburbs around a
11 light or commuter rail station, and by offering several apartment sizes and price ranges, one may
12 meet the needs and suit the preferences of many households – including families with children,
13 those desiring more than one parking space, and, at the same time, single persons of relatively
14 low income but who would value the transit access and relatively affordable accommodation.

15
16 The previous examples are just some of the applications one might devise from the results of this
17 work. The data set and various models are hoped to be a valuable source for more informed
18 policymaking, land development practices, and transportation system design.
19

20 **6 Conclusions and Extensions**

21 This work provides new insights into location and dwelling choices by those living in apartments
22 in the Austin area. One particularly valuable aspect of the research lies in the data set itself. The
23 focus is on apartment dwellers (rather than home owners), and questions range from reasons for
24 moving, to rent and apartment attributes, to tradeoffs between pairs of key access-dwelling
25 qualities, and to ratings of individual attributes.
26

27 One finds that apartment dwellers may have very different reasons for moving than home owners
28 and others; for example, a new job (or job transfer) is far more common. Rent and apartment size
29 models reveal several tradeoffs that households make: for example, another bedroom adds
30 approximately \$119 to monthly rent and newness \$44, while access to commercial centers adds
31 around \$24. Rents fall by about \$20 per month for each additional mile away from the CBD, and
32 by \$24 for each added commute-time minute. A higher bus stop density also is associated with
33 lower rents. Urban area apartments run about 75 square feet smaller than others, ceteris paribus,
34 and those in more densely populated neighborhoods run smaller (about 28 square feet smaller for
35 every added person-per-acre).
36

37 Binary logit models of stated preferences suggest that multi-person households, married couples,
38 and those with children tend to prefer larger and newer apartments as well as better recreation
39 facilities and suburban locations, while single-person households are more likely to choose a
40 shorter commute and more central locations. Additionally, the results suggest that women prefer
41 more space to a percentage reduction in commute time, as compared to men. Women and non-
42 Caucasian apartment dwellers tend to be more concerned with accessibility. Those living without
43 children tend to more concerned about commute times and freeway access, everything else
44 constant.
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46 Finally, although this study offers significant insights, several extensions would be valuable.
47 Ideally, more persons in more locations would be surveyed, producing greater variety in spatial
48 as well as demographic characteristics. A random sample (rather than choice-based sample) of
49 apartment dwellers would permit calibration of a location choice model, to more formally
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3 determine the neighborhood, price, and access factors (and tradeoffs) that are at play in
4 apartment choice. With such data sets and models on hand, prediction of future land use patterns
5 as well as the viability of new forms of residential design will be greatly enhanced.
6

7 **Acknowledgments**

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10 of Ahmed Qatan, Shadi Hakimi, Nick Lownes, and Shashank Gadda, the data set would not have
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12 process, as well as Stacey Bricka. A special thanks to Ahmed Qatan for making the survey
13 available on the Internet. We would also like to recognize the Southwest University
14 Transportation Center (SWUTC) for funding this research project.
15

16 **Endnotes**

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18 ¹ This sampling area is the 787xx Zip Code Tabulation Area (ZCTA), which has a population of 777,789.

19 ² This technique uses a stochastic draw to impute the data, by adding a random term to a regression models estimate
20 of age. Little and Rubin (1987) concluded that this method suffers less from bias than relying on the regression
21 model's "best" or average guess. The two-sample test used data from records providing age information, and those
22 without.
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Table 1. Characteristics of sample

Variables		Min.	Max.	Mean	Std. dev.	Number of observations
Apartment features (self-reported)	Number of bedrooms	1	4	1.61	0.65	235
	Number of bathrooms	1	4	1.46	0.56	234
	Rent (dollars per month)	150	1,500	673.33	263.54	240
	Interior size (square feet)	300	1,700	861.70	285.36	235
	Commute to work/school	3	100	19.59	14.34	222
	Travel time to grocery store	3	100	8.01	8.90	232
	Travel time to mall	3	100	15.86	12.86	238
Household information	Household size	1	4	2.08	1.03	240
	Number of workers in household	0	4	1.28	0.80	239
	Number of children	0	4	0.48	0.93	239
	Number of licensed drivers in household	0	4	1.52	0.79	238
	Number of vehicles	0	5	1.38	0.76	240
	Household income (\$1000/year)	13	200	37.86	27.95	240
Respondent information	Married	0	1	0.28	0.45	237
	Age	18	83	32.83	12.80	240
	Male (indicator)	0	1	0.51	0.50	240
	Number of days per week typically driven	0	7	5.42	2.30	238
	Caucasian	0	1	0.48	0.50	239
	Hispanic/Latino	0	1	0.28	0.45	239
	African-American	0	1	0.10	0.31	239
	Asian	0	1	0.09	0.29	239
	Other ethnicity	0	1	0.04	0.20	239
	Non-Caucasian	0	1	0.52	0.50	239
	Living alone	0	1	0.37	0.48	240
	Living with friends	0	1	0.15	0.36	240
	Living with family	0	1	0.29	0.46	240
	Living with significant other	0	1	0.17	0.38	240
	Less than high school	0	1	0.05	0.23	238
	High school	0	1	0.37	0.48	238
	Associate's or technical degree	0	1	0.16	0.37	238
	Bachelor's degree	0	1	0.29	0.46	238
	Master's degree or higher	0	1	0.13	0.33	238
	Employed full-time	0	1	0.56	0.50	238
Employed part-time	0	1	0.09	0.29	238	
Full-time student	0	1	0.19	0.40	238	
Homemaker	0	1	0.03	0.16	238	
Unemployed	0	1	0.08	0.27	238	
Retired	0	1	0.05	0.21	238	
Supplementary data)	Urban (indicator)	0	1	0.74	0.44	240
	Distance to CBD	1	15	6.59	2.95	240
	Neighborhood mean travel time to work	17	27	22.90	2.88	240
	Neighborhood median household income	17,596	63,662	34,542	13,044	240
	Neighborhood median rent	581	911	714.62	86.18	240
	Cost for home-based work trips	4,477	6,998	4,992	697	240
	Cost for home-based non-work trips	4,671	7,718	5,291	825	240
	Population density (people/ square mile)	900	11,437	3,366.58	1,888.19	240
	Percent of non-Caucasian residents	0.12	0.64	0.35	0.16	240
	Employment per square mile	212	6,821	1,551.52	1,530.36	240
Bus stops per square mile	11	150	71.26	36.63	240	

Table 2. Primary reason for moving

Primary Reason for Moving (Sample Results)	Frequency	Percent*	Primary Reason for Moving (Census Results)	Percent
Wanted new/better apartment	44	18.74%	New/better house/apartment	19.8%
Easier commute	40	17.03%	Other family reason	12.6%
Other	36	15.33%	Other housing reason	11.0%
New job/job transfer	32	13.57%	Wanted to own home/not rent	10.2%
Wanted/needed less expensive housing	24	10.33%	New job/job transfer	8.8%
Planned to attend or graduate from college	15	6.33%	To establish own household	7.0%
Marriage or divorce	14	6.16%	Change in marital status	6.7%
Wanted to rent	13	5.64%	Cheaper housing	6.5%
Birth/adoption	9	3.73%	Better neighborhood/less crime	3.8%
Change of climate	6	2.40%	Closer to work/easier commute	3.2%
Retiring	1	0.39%	Attend/leave college	2.5%
Health reasons	1	0.36%	Other reason	2.5%
* corrected percentages weighted for Austin’s apartment dwelling population			To look for work/lost job	1.9%
			Other work reason	1.4%
			Health reasons	1.4%
			Change of climate	0.4%
			Retired	0.3%

Table 3. Mean rank of importance of housing and location attributes

Housing/Location Attributes	Mean Rank (where 1 is very unimportant and 5 is very important)
Price	3.663
Commute time to work	3.277
Perception of crime rate	3.246
Attractive neighborhood appearance	3.166
Commute time to school	3.145
Access to major freeways	3.095
Noise	2.991
Distance/travel time to shopping	2.645
Social composition of the neighborhood	2.632
Neighborhood amenities / recreational facilities	2.621
Access to public transportation	2.571
Views	2.494
Closeness to friends or relatives	2.406
Quality of local public schools	2.243
Distance to local public schools	2.218

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Table 4. Final linear regression models of rent and square footage

Variables		Monthly rent (\$)		Square footage (sq. ft.)	
		β	<i>p</i> -value	β	<i>p</i> -value
	Constant	993.71	0.00	809.98	0.00
Apartment and neighborhood features (self-reported)	Number of bedrooms	118.90	0.00	152.52	0.00
	Number of bathrooms	109.28	0.00	179.17	0.00
	Commercial center within walking distance (0-4)	24.41	0.13		
	Relatively new apartment (0-4)	43.81	0.00		
	Travel time to mall (min.)			-2.60	0.02
Age and ethnicity	Age				
	Non-Caucasian	-52.67	0.06		
Education level	Lower education (base)	0	N/A	0	N/A
	Master's degree or higher	110.01	0.00	94.29	0.02
Household information	Number of children	-46.57	0.00	-22.55	0.16
	Household income (per \$1000 annual salary)	0.81	0.19	0.77	0.16
Supplementary data	Urban Indicator			-70.06	0.09
	Distance to CBD (miles)	-19.50	0.01	-37.50	0.00
	Neighborhood mean travel time to work (minutes)	-24.31	0.00		
	Population density (people per square mile)	0.02	0.03	-0.04	0.00
	Number of bus stops per square mile	-2.52	0.00		
	Number of observations	209		229	
	Adjusted R ²	0.551		0.508	

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Table 5. Final binary logit models of stated preference questions

Variables		Scenario 1: 200 extra sq. feet vs. shorter commute		Scenario 2: Friends/relative s nearby vs. light rail to work		Scenario 3: Suburban location vs. downtown with one parking spot (extra spot = \$60)		Scenario 4: Closer to shopping vs. larger kitchen		Scenario 5: Close to bus stop vs. view of park		Scenario 6: Brand new complex vs. 5 miles to shopping center	
		β	p	β	p	β	p	β	P	β	p	β	p
	Constant	-0.64	0.73	3.66	0.00	1.60	0.00	-2.81	0.00	-2.31	0.02	4.69	0.03
Living situation	Number of workers			-0.55	0.01					-0.25	0.12		
	Number of children												
	Married	0.80	0.02			0.68	0.09	0.42	0.16				
	Living alone	-0.68	0.04	-0.78	0.02	-0.99	0.00					-0.59	0.09
Ethnicity and gender	Non-white					0.74	0.02	0.64	0.02	0.73	0.01	-0.64	0.06
	Male	-0.70	0.02			-0.61	0.05						
Education	Less than high school							-1.38	-0.08				
	Master's or higher					1.43	0.01						
Employment Status	Full-time	0.51	0.09										
	Retired											-1.52	0.06
Income	Household income(per \$1000 annual salary)									-0.01	0.01	0.01	0.08
Supplementary data	Urban indicator	-0.78	0.13	-1.47	0.00	-1.19	0.00	0.92	0.01	0.84	0.04	-0.93	0.07
	Distance to CBD	-0.18	0.01	-0.21	0.00					0.14	0.05		
	Neighborhood mean travel time to work	0.12	0.07									-0.11	0.15
	Population density									0.00	0.02		
	#Observations	231		235		233		233		236		235	
	Log likelihood	-143.08		-150.41		-127.54		-148.23		-146.21		-117.14	
	Adjusted rho square	0.073		0.050		0.181		0.057		0.074		0.256	
	Market shares (apt. 1 vs. apt. 2)	47% vs. 53%		55% vs. 45%		66% vs. 34%		41% vs. 59%		46% vs. 54%		77% vs. 23%	

Table 6. Final ordered probit models of importance of commute, distance/travel time to shopping, access to public transportation, and access to major freeway(s)

Variables		Commute time		Distance/ travel time to shopping		Access to public transportation		Access to major freeway(s)	
		β	<i>p</i>	β	<i>p</i>	β	<i>p</i>	β	<i>p</i>
Constant		1.950	0.000	1.046	0.000	2.224	0.000	1.502	0.000
Household/ respondent information	Number of workers in household			-0.254	0.006	-0.260	0.017		
	Presence of at least one child in household	-0.941	0.000					-0.370	0.129
	Married and have at least one child			-0.698	0.043				
	Age			0.016	0.007				
	Male	-0.372	0.021					-0.277	0.072
	Number of vehicles available in household					-0.311	0.017		
	Household income (per \$1000 annual salary)					-9.45E-03	0.001		
	Full-time student			-0.260	0.128	0.338	0.048		
	Hispanic/Latino			0.327	0.056			0.369	0.046
	African-American							1.048	0.000
Non-Caucasian	0.569	0.000			0.399	0.006			
Living situation	Living alone					-0.695	0.000		
	Living with friends								
	Living with family			0.505	0.002			0.435	0.013
	Living with significant other			0.387	0.059			0.419	0.021
Highest level of education	Less than high school							0.263 ¹	0.085
	High school								
	Associate's or technical degree								
	Bachelor's degree					-0.389 ²	0.019		
Master's degree or higher	0.805	0.000							
Thresholds	μ (0)	0	N/A	0	N/A	0	N/A	0	N/A
	μ (1)	0.773	0.000	1.270	0.000	0.681	0.000	0.854	0.000
	μ (2)	2.193	0.000	2.517	0.000	1.655	0.000	2.260	0.000
#Observations		221		224		214		228	
Loglikelihood		-216.293		-263.968		-264.570		-244.230	
Log Lik: constants only		-238.425		-283.049		-293.738		-260.351	
Adjusted LRI		0.062		0.028		0.077		0.049	

¹ Represent a combination variable of less than high school or high school education level.

² Represent a combination variable of associate's degree, bachelor's degree, or master's degree or higher education level.