

1 **How Far to Live and with Whom? The Role of Modal Accessibility on Student's Choice**  
2 **of Living Arrangements and the Distance they are Willing to Live from University in**  
3 **Toronto**

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7  
8 Brittany Chung, BAsC  
9 MAsC Candidate  
10 Department of Civil Engineering  
11 University of Toronto  
12 Email: [brittany.chung@mail.utoronto.ca](mailto:brittany.chung@mail.utoronto.ca)  
13

14  
15 Md Sami Hasnine, MAsC  
16 PhD Candidate  
17 Department of Civil Engineering  
18 University of Toronto  
19 Email: [sami.hasnine@mail.utoronto.ca](mailto:sami.hasnine@mail.utoronto.ca)  
20

21  
22 Khandker Nurul Habib, Ph.D., PEng (\*Corresponding Author)  
23 Associate Professor  
24 Department of Civil Engineering  
25 University of Toronto  
26 Email: [khandker.nurulhabib@utoronto.ca](mailto:khandker.nurulhabib@utoronto.ca)  
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41

1 **Abstract**

2 This study investigates the factors influencing university student’s living arrangement  
3 choices and the distance they are willing to live from their university. Particular attention  
4 is given to the effects on these choices by student’s typical commuting mode choice to and  
5 from campus. The dataset used for the study was collected by a travel survey on  
6 postsecondary students across four universities in Toronto. An econometric model is  
7 developed to jointly model three choices: student’s typical commuting mode choice, their  
8 discrete living arrangement choice, and the distance they are willing to live from their  
9 campus. The model reveals that the students who are likely to leave the family home for  
10 university, also have the tendency to live close to the university. Modal accessibility was  
11 found to plays a crucial role in the trade-off between students’ choices of leaving their  
12 family home and the distance they are willing to live from the university. Choice of staying  
13 in their family home is positively affected by larger household sizes and the presence of  
14 senior members (age 75+). Walk and cycling accessibility appear to be the most influential  
15 factor in the distance students are willingness to live away from their university. Cost of  
16 housing, neighborhood environment, proximity to transit, and proximity to friends/family  
17 also have significant effects on the influence of modal accessibility in defining student’s  
18 residence distance from their university. These suggest that affordable housing choices by  
19 the university campuses can significantly increase the number of active mode users among  
20 the student population in Toronto.

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## 1. Introduction

Understanding living arrangement and resident location choices are key interests of urban planners and policy makers as these have significant impacts on various urban dynamics. However, predicting these can be challenging as there are many complex interdependencies. For example, employment opportunities and accessibility to the transportation alternatives can motivate individual's resident location and living arrangement choices. In turn, residential demand in an area can also influence the surrounding area's employment and transportation structure. Resident location choice also involves various behavioural processes and decisions such as when to move, where to move, and who to live with. Studies have shown that location choice behaviour and relocation probability changes over an individual's life span (Beige and Axhausen 2008; Kim et al 2005; Schirmer et al 2014). These changes are often induced by major milestone events such as getting married, a change in place of employment, the start of a postsecondary education, etc. However, the literature on postsecondary student's travel behaviour and resident location choice have been small in number. This is largely credited to the lack of data available on this niche cohort. In fact, they are often under represented in regional travel studies (Hasnine et al. 2017). For example, in a large scale (5% household sample) travel survey in southern Ontario, postsecondary students constituted only 2.5% of the sample but in reality, they contribute 23% of the area's population (Habib et al. 2017).

One data collection initiative that attempted to better understand postsecondary students' travel behaviour was a multi-university student travel survey in Toronto, the StudentMoveTO (2015). This was a joint effort of four universities with seven campuses in Toronto Area to better understand their students' travel behaviour through a one day travel diary survey conducted in the fall of 2015. This paper makes use of this survey data set to investigate factors influencing university students' living arrangement choice and the distance they are willing to live from the university in relation to the perceived accessibility of various commuting mode choice alternatives. An econometric model is developed to jointly estimate these three choices. Four exclusive living arrangement choices are modelled: living with parents/family, living with a partner, living alone, and living with roommates. For the subset of students who leave their parental home, a resident location choice model is developed to estimate the distance students are willing to live from campus. A travel mode choice model for a typical fall-day is used to compute modal accessibility perceptions which was used as inputs in the living arrangement and residence distance location models.

The remainder of the paper is organized as follows. The first section is a summary of the relevant literature on postsecondary student's mode choice, residence distance location choice, and living arrangement. The paper then discusses the data set used for empirical

1 investigation and proceeds to explain the formulation of the econometric model. Next, the  
2 results of the model are presented along with its policy relevancies. Finally, the paper  
3 concludes with a summary and recommendations for future research.

## 4 5 **2. Literature Review**

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7 Studies show that universities can significantly affect the demographic, economic and  
8 housing characteristics of its surrounding neighbourhoods (Cortes 2004; Cisneros 1996;  
9 Macintyre 2003). Residential location choice models typically ignore these effects and  
10 often cluster the students' residence choice behaviour with the general population  
11 (Schirmer et al 2014). However, researchers found that this niche population can exhibit  
12 considerably different behaviour from that of the general population (Simons et al. 2017;  
13 Zheng et al. 2009). The neglect of this demographic cohort is largely attributed to the  
14 frequent under representation of postsecondary student in regional travel surveys  
15 (Hasnine et al. 2017). Urban planners and researchers recognize this shortcoming and thus  
16 the interest in studying postsecondary student travel behaviour has been gaining some  
17 traction in recent years.

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19 Among the few studies on university students' travel behavior, several investigate the  
20 mode choice and general travel patterns of students. Overall, many of the mode choice  
21 studies conclude that postsecondary students tend to use more active modes of transport,  
22 and make more home-based university trips particularly during off-peak periods (Khattak  
23 et al. 2011; Wang et al 2012; Zheng et al. 2009; Simons et al. 2017). A survey of four major  
24 universities in Virginia showed students living off campus are more likely to drive (Khattak  
25 et al. 2011; Wang et al 2012). A similar survey effort, but in the context of university  
26 students in Toronto, supports these findings. Hasnine et al. (2017) developed a series of  
27 discrete choice models to capture the commuting mode choice behaviour of Toronto's  
28 university students and found students travelling longer distances to their school opted for  
29 auto-drive and auto-passenger modes. Transit and walk access mode users particularly  
30 lived in close proximity to subway stations and highly transit accessible areas (Hasnine et  
31 al. 2017). In terms of walking commutes, a study by Manaugh and El-Geneidy (2013)  
32 investigated the impact of walking satisfaction on housing choice of students at McGill  
33 University in Montreal, Quebec. Their statistical cluster analysis revealed that persons who  
34 valued the environment and exercise were likely to choose a home location that requires a  
35 longer walk to their destination and were less likely to base their decision on proximity  
36 compared to individuals that valued.

37  
38 It is obvious that the mode choice of students is directly related to the commuting distance  
39 between the university and students' residence. However, the body of literature on  
40 students' residence location choice is small in number. Some researchers have attempted  
41 to investigate postsecondary school choice based on the distance between the schools and

1 the student's parental homes. Many of these studies reveal that distance from home to  
2 school is a key factor in students' university choice (Frenette 2002; Frenette 2003; Pinjari  
3 et al. 2011; Card 1993). In Frenette's (2002) econometric analysis of postsecondary  
4 participation in Canada, it is found that prospective students living between 40km and  
5 80km from a university are only 69% as likely to attend as students who live closer to the  
6 university.

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8 Since commute time is negatively related to residential location choice, Pinjari et al. (2011)  
9 showed that individuals generally try to locate within proximity to their work/school.  
10 However, their study does realize the tendency to live closer is less pronounced in higher  
11 income households compared to lower income households, possibly because higher income  
12 households can afford the higher transportation costs (Pinjari et al. 2011). It is evident  
13 from many of other studies that income exerts a strong influence on university  
14 participation rates. A follow-up study by Frenette (2003) found Canadian students from  
15 lower and middle-income families are most negatively affected by living far away from  
16 universities. Particularly for students from lower income families, students living beyond  
17 commuting distance to a university are less likely to attend university (Frenette 2003;  
18 Christofides et al. 2001). Those that live close to postsecondary institutions have relatively  
19 higher education and higher income (Card 1993).

20  
21 In terms of living arrangements, student's decision of continuing to live in their parental  
22 home or moving closer to the university has been rarely studied. One of the more relevant  
23 studies was done by Sà et al. (2012) which estimated Dutch students' university choice and  
24 decision on living arrangements using a nested logit model. It revealed that the two  
25 decisions are simultaneously linked and distance deters both students living in their  
26 parental home and those choosing to live elsewhere (Sà et al. 2012). Logistic regression  
27 models developed by Mulder and Clark (2002) showed that the likelihood of US college  
28 students leaving their parental home increases with parental income and is positively  
29 affected by the father's education level. Based on the reviewed literature, most studies  
30 define living arrangements as binary choices: stay in the parental home, or leave parental  
31 home. Many ignore the different possible student living arrangement outside the parental  
32 home, such as living alone, with a partner, or with roommates which are investigated in  
33 this paper.

34  
35 As discussed, there have been notable attempts to model university student's mode choice  
36 accessibility, residential location, and living arrangements. Though many studies in the  
37 reviewed literature allude to the idea that these choices are interconnected in some form,  
38 no study has attempted to jointly investigate these three choices. Therefore, this paper  
39 contributes to the literature by providing a comprehensive, holistic view of the interactions  
40 between the three decisions. The joint models capture the key factors influencing the

1 important transitional period for university students and their impact on Toronto's urban  
2 environment.

3

### 4 **3. Data for Empirical Investigation**

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6 The data used in this study were retrieved from the StudentMoveTO web-survey, which is a  
7 joint data collection effort initiated by four Toronto universities: Ontario College of Art  
8 and Design (OCAD) University, Ryerson University, York University, and the University of  
9 Toronto. Collectively, the four universities have seven campuses across the city of Toronto  
10 with a total enrollment of over 184,000 postsecondary students. In the fall of 2015, survey  
11 invitations were sent to the university email address of all current students in the seven  
12 campuses. The data collection initiative collected 15,226 survey responses which yield an  
13 8.0% completion rate.

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15 The information collected includes socio-economic, personal and household attributes, as  
16 well as a one-day travel diary of their last commuting day to school. In addition, students  
17 were asked several questions on their living arrangement, housing characteristics, and  
18 commuting patterns. Table 1 provides the summary statistics of select key variables in the  
19 sample dataset. Although the mode share values reported are for typical trips in the fall,  
20 more than 90% of the students claim to use the same travel modes in the winter. It is  
21 evident that majority of students typically use transit to commute between home and  
22 school (64%). Active-modes occupy a significant percentage of the mode share, as 19% of  
23 respondents walk and 9% cycle. Interestingly, two-thirds of the respondents own a car and  
24 possesses a driver's license, though drivers occupy only 6% of the mode share.

25

26 To better understand these mode choices, attitudinal questions such as what are the key  
27 reasons for choosing their typical commute mode were asked in the survey. Over 50% of  
28 the transit users chose transit because it is perceived as one of the most convenient and  
29 least expensive options. 20% of these transit users also claimed that they would walk or  
30 cycle, though they live too far for those to be realistic options. On a similar note, 80% of  
31 walkers chose to walk because of their proximity to campus and it is free of cost. Also, 80%  
32 of cyclists find cycling to be one of the faster and least expensive means of transportation.  
33 On the other hand, students who use driving as a commuting mode find that driving is their  
34 fastest option, and they live too far for other means of transportation to be practical. The  
35 small percentage of rideshare commuters state that it saves time and parking expenses,  
36 and it is convenient since their carpool partner lives near by. Based on these responses, it is  
37 apparent that students' mode choices are highly sensitive to travel time/distance and costs.  
38 Sensitivity to costs is understandable as it appears the majority of these students live in  
39 low-income households. In terms of distance, the distribution of the distance between  
40 student's homes to their main campus appears to skew to the left. On average, their

1 commuting distance is 12.5km but has an upper limit of approximately 70km. The left skew  
 2 explains the substantial share of active mode users, as well as transit users.

3

4 **Table 1. Summary statistics of select variables**

<b>Discrete Variables</b>				
<i>Variables</i>	<i>% Share</i>	<i>Variables</i>	<i>% Share</i>	
<b>Education level</b>		<b>Household Income Level</b>		
Undergraduate	70%	Less than \$ 30,000	35%	
Graduate	28%	\$ 30,000 - 59,999	22%	
Other	2%	\$ 60,000 - 89,999	16%	
<b>Student Status</b>		\$ 90,000 - 119,999	11%	
Full Time	90%	\$ 120,000 - 149,999	16%	
Part Time	8%	\$ 120,000 - 149,999	6%	
Other	2%	\$ 150,000 - 179,999	4%	
<b>Gender</b>		\$ 180,000 +	6%	
Female	64%	<b>Building Type</b>		
Male	36%	Apartment or Condo	43%	
<b>Car ownership</b>		Detached house	32%	
Own car	68%	Semi-detached house	14%	
Do not own car	32%	Row house	7%	
<b>Driver License Ownership</b>		On-campus residence	4%	
Possess a driver license	63%	<b>Household Ownership</b>		
Does not possess a driver license	37%	Owner	47%	
<b>Main mode of transportation in the fall</b>		Tenant	53%	
Transit	64%	<b>Living arrangement</b>		
Walk	19%	Live with family/parents	53%	
Bike	9%	Live with roommate	18%	
Driving	6%	Live with partner	17%	
Rideshare	2%	Live alone	12%	
<b>Continuous Variables</b>				
<i>Variables</i>	<i>Mean</i>	<i>St. Dev.</i>	<i>Maximum</i>	<i>Minimum</i>
Distance from home to main campus (km)	12.5	11.8	67.3	0.3
Number of vehicles in household	1.0	1.0	9.0	0.0
Number of years living in current home location	5.6	6.6	50.0	0.0
Household Size	3.2	1.6	20.0	1.0
Age of respondent	24.2	6.9	64.0	16.0
Household Average Age	30.9	8.6	68.0	3.5
Number of dependent children in the family	0.3	0.8	8.0	0.0
Number of seniors (75+) in the family	0.0	0.2	3.0	0.0

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1 The student's sensitivity to cost is also reflected in their residence location choices.  
2 Approximately 30% of respondents claimed that the cost of housing was the primary factor  
3 influencing their choice of residence. This response corresponds to 53% of respondents  
4 living at their parental home since it is the least expensive housing option. 16% of the  
5 residence choices are supposedly based on the convenience of walking or biking to campus.  
6 Interestingly, 17% of respondents stated that their residence location choice was beyond  
7 their control.

8  
9 In this study, the transportation level-of-service variables (travel time, cost, etc.) for  
10 residence to the university is generated by using a regional planning model (traffic  
11 assignment component), which is calibrated by using the large regional household travel  
12 survey data and is used by the City of Toronto for its planning exercises (Data Management  
13 Group 2011).

14

#### 15 **4. Econometric Model**

16

17 We modelled three choices jointly and these are:

18

- 19 1. Choice of living arrangements: living with parents/family; living apart from the  
20 parents/family, but with a partner, living with a roommate(s), and living alone.
- 21 2. Regular choice of mode for a home to the University trips: driving a car, car passenger  
22 (carpooling), public transit, bicycle and walk.
- 23 3. Distance willing to live from the university for those not living with the parents/family.

24

25 The dataset does not provide any information on alternative residence location choices and  
26 generating such alternatives would require many strict assumptions that may affect the  
27 results of the analysis. So, we proceed to consider the distance between the university and  
28 the residence as the continuous variable of choice and modelled it by using direct utility  
29 maximization approach. Finally, three alternative choices are investigated and these are  
30 discrete living arrangement choice, discrete mode choice, and students' willingness to live  
31 from the university which is quantified by a continuous distance variable. It is  
32 understandable that these three are correlated choices. We propose that the choices of  
33 living arrangement and residence distance are correlated directly through modal  
34 accessibility (derived from mode choice) and indirectly through correlated random utility  
35 components. Figure 1 depicts the simultaneity of the three choice dimensions under  
36 investigation.

37

38 Considering random utility-based modelling of choice makings, the utility functions (U) of  
39 the three choices are:

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$$41 \quad U_m = (\sum \beta y)_m + \varepsilon_m; \quad m=1, 2, 3, 4 \text{ and } 5 \quad (1)$$

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 2  $U_R = (\sum \alpha x)_R + \varepsilon_R ; R=1, 2, 3, \text{ and } 4$  (2)

3  
 4  $U_l = (\sum \gamma z)_l + \varepsilon_l ; l=1, 2 \text{ and } 3$  (3)

5  
 6 Here,  $m$  indicates mode choice,  $R$  indicates living arrangement choice, and  $l$  indicates  
 7 distance willing to love from the university.

8 Alternative values of  $m$ ,  $R$  and  $l$  are indicated in Figure 1.

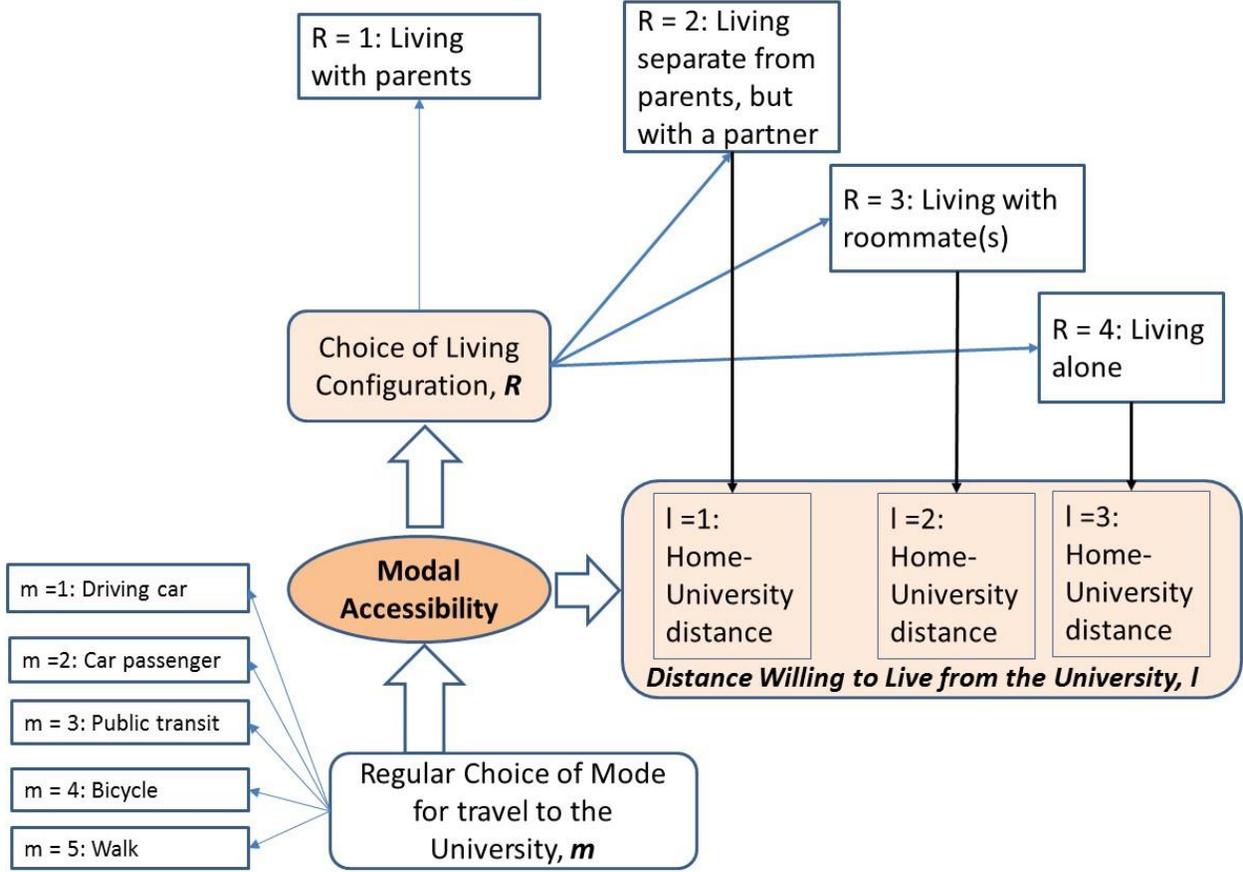
9  $\varepsilon$  indicates the random components of the utility functions

10  $y$  indicates the systematic attributes that explain the choices of alternative modes  
 11 with  $\beta$  as corresponding coefficients

12  $x$  indicates the systematic variables explaining the choices of living arrangement  
 13 choice with  $\alpha$  as corresponding coefficients

14  $z$  indicates the systematic variables explaining the choice of resident location  
 15 defining the distance between residence and the university with  $\gamma$  as corresponding  
 16 coefficients

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 19 **Figure 1: Modelling the influence of modal accessibility on choices of living**  
 20 **arrangement and the location of the residence**

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Equation 1 and 2 are of discrete choices and so the corresponding utility functions are of indirect utility functions, but equation 3 is of continuous distance from the university to the resident location and of a direct utility function. Assumptions of an Independent and Identically Distributed (IID) Type I extreme distribution with unit scale assumptions for  $\varepsilon_m$  lead to standard multinomial logit models. Thus resulting to the following mode choice model:

$$\Pr(m) = \frac{\exp((\sum \beta y)_m)}{\sum_{m'=1}^5 \exp((\sum \beta y)_{m'})} \quad (4)$$

As per the properties of RUM and the distributional assumption of  $\varepsilon_m$ , the expected maximum utility of mode choice is (EMU):

$$EMU = \ln(\sum_{m'=1}^5 (\sum \beta y)_{m'}) \quad (5)$$

This is also known as modal accessibility (Akbari and Habib 2015). Modal accessibility is a measure of ease of getting access to a particular destination by available modes. The value of such modal accessibility increases with increasing number of alternative feasible modes and decreases with increasing travel time to get to the destination. So, the EMU is a measure of the performances of the available modes and so we consider that this has systematic effects on both choices of resident configuration choice and residence distance location choice. Considering this assumption and an IID Type I Extreme Value distribution with unit scale for  $\varepsilon_R$  the choice probability of resident choice becomes:

$$\Pr(R) = \frac{\exp((\sum \alpha x + EMU)_R)}{\sum_{R'=1}^4 \exp((\sum \alpha x + EMU)_{R'})} \quad (4)$$

For the continuous variable,  $l$ , considering a log-normal<sup>1</sup> distribution with  $\sigma^2$  variance for  $\varepsilon_l$  and a Cobb-Douglas direct utility maximization formulation, the probability of a chosen distance from the university becomes:

$$\Pr(l) = \frac{1}{\sigma} \phi \left( \frac{\ln(l) - ((\sum \gamma z)_l + EMU)}{\sigma} \right) \quad (5)$$

It is also understandable resident configuration and the location choices can be correlated. In the above formulations of resident configuration and location choices, modal accessibility plays a common moderating role linking the systematic utility components of both choices. Such link connects observed systematic components of the two choices, but there could be unobserved correlated factors that influence both choices too. A further

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<sup>1</sup> Lognormal distribution assumption helps maintain the fact that the distance must be non-negative.

1 assumption on correlated observed factors influencing discrete resident configuration choice  
 2 and the continuous distance between the university and the resident choice can be induced  
 3 by following Lee's transformation technique (Lee 1983; Habib and Hasnine 2017), which is  
 4 also known as Gaussian Copula approach. Under this approach, the discrete marginal  
 5 probabilities of resident configuration choice as specified in equation 4 is transformed into  
 6 an equivalent standard normal random variable. This transformed normal variable is  
 7 assumed to be jointly distributed with the normal random error  $\varepsilon_l$  as per a bivariate normal  
 8 distribution form. The resulting joint probability of resident configuration choice and the  
 9 residence distance location choice from the university becomes:

$$11 \quad P(R\&l) = \frac{1}{\sigma} \phi \left( \frac{\ln(l) - ((\sum \gamma z)_l + EMU)}{\sigma} \right) \Phi_2 \left( \Phi^{-1} \left( \frac{\exp((\sum \alpha x + EMU)_R)}{\sum_{R=1}^4 \exp((\sum \alpha x + EMU)_{R'})} \right) - \rho \left( \frac{\ln(l) - ((\sum \gamma z)_l + EMU)}{\sigma} \right) \right)$$

12  
13 (6)

14 Here,  $\phi(\cdot)$  is the probability density function of a univariate standard normal distribution  
 15  $\Phi^{-1}(\cdot)$  is the inverse of the cumulative density function of a univariate standard  
 16 normal distribution  
 17  $\Phi_2(\cdot)$  is the cumulative density function of a bivariate standard normal distribution  
 18  $\rho$  is the correlation coefficient between the unobserved factors influencing the  
 19 choices of  $R$  and  $l$ , which varies between -1 and +1

20  
 21 Under this formulation, the likelihood function (L) of an observation (observed travel mode  
 22  $m$ ; resident configuration type  $R$  and distance  $l$ ) becomes:

$$24 \quad L = \prod_{R=1}^4 \prod_{m=1}^5 \left( \frac{1}{\sigma} \phi \left( \frac{\ln(l) - ((\sum \gamma z)_l + EMU)}{\sigma} \right) \Phi_2 \left( \Phi^{-1} \left( \frac{\exp((\sum \alpha x + EMU)_R)}{\sum_{R=1}^4 \exp((\sum \alpha x + EMU)_{R'})} \right) - \rho \left( \frac{\ln(l) - ((\sum \gamma z)_l + EMU)}{\sigma} \right) \right) \right)^{\delta_R}$$

$$25 \quad \left( \frac{\exp((\sum \beta y)_m)}{\sum_{m'=1}^5 \exp((\sum \beta y)_{m'})} \right)^{\delta_m}$$

26  
27 (7)

28 Here,

29  $\delta_R$  is an indicator variable for resident type choice  $R$  taking the value 1 for type  $R$   
 and 0 otherwise

30  $\delta_m$  is an indicator variable for mode choice  $m$  taking the value 1 for type  $m$  and 0  
 31 otherwise

1  
2 Equation (7) presents a closed form likelihood function that can be estimated by using the  
3 classical maximum likelihood estimation approach. In this paper, we programmed the  
4 likelihood function in GAUSS and used its MAXLIK estimation routine for the maximum  
5 likelihood estimation approach (Aptech 2017).  
6

## 7 **5. Empirical Investigation and Policy Relevance**

8

9 The empirical model is presented in Table 2. The joint model estimates student's typical  
10 commuting mode choice to and from university, their living arrangement choice and the  
11 distance they are willing to live from the university. We considered a 95 percent confidence  
12 limit with expected signs to determine the variables in the model. The final model has a  
13 total of 57 parameters. We retained some parameters with lower than 95 percent  
14 confidence as those provide behavioural insights. The model fitness is compared against a  
15 null model that considered all discrete choices as equally likely and only constants for the  
16 continuous distance model component. The adjusted Rho-squared value of the final model  
17 is 0.23, which is a reasonably good fit considering the complexity and multiple choices  
18 modelled jointly.  
19

20 The correlation between unobserved factors affecting the choice of living configuration and  
21 distance willing to live from the university campus is significant and negative. This  
22 indicates that the unobserved variable that influence the choice of not staying in  
23 family/parental house also recued the willingness to live far from the university campus.  
24 This is intuitive and thus an implicit validation of the captured relationship of the three  
25 choices investigated in this empirical model. It refers to the fact that students who leave  
26 their family home to attend the university will obviously tend to live close the university  
27 campus. However, the fact of the matter is that there are many systematic variables/factors  
28 that defy this inherent and implicit choice by allowing students to choose one of the three  
29 living configuration choices (except staying in family/parental home) and corresponding  
30 residence location that may not be optimal in terms of proximity to campus. In this context,  
31 the effects of the investigated systematic variable captured in the model give us clear  
32 understandings on the choice constraints of postsecondary student living in big cities, such  
33 as Toronto.  
34

35 The joint model used a simplified mode choice model to make the modal accessibility  
36 variable endogenous to the living arrangement and distance choice model as it is an input  
37 of these model components. The mode choice model has alternative-specific-constants,  
38 travel time and travel cost. This gives a subjective value of travel time savings of  
39 approximately \$4 per hour, which is a conservative estimate given that this is of regular fall  
40 travels. This value can vary for day-to-day trips and with seasons. Since living arrangement  
41 and location choices are of longer-term choices than day-to-day trip choices, the mode

1 choice model representing general fall university trips are considered as opposed to  
2 specific school trip mode choice.

3

4 The resulting model reveals that the expected maximum utility of mode choice to be  
5 statistically significant, thus confirming our speculation that the choice of living  
6 arrangement and distance students are willing to live from the university are directly  
7 correlated with modal accessibility. The group that appears to be most sensitive to modal  
8 accessibility is the students living with roommates. Those living with roommates tend to  
9 have a full-time student status and are relatively younger in age. Though individuals in this  
10 group are rather likely to own a driving license, they are unlikely to have personal vehicles  
11 at their residences. This discrepancy between vehicle ownership and license ownership is  
12 explained by the group's low-income level. The living arrangement model shows that those  
13 living with roommates tend to have an annual household income less than \$30,000,  
14 therefore making many of these students financially unfit to own a vehicle. Notably, the  
15 residence distance location model suggests that younger individuals choosing to live with  
16 roommates are more inclined to live closer to campus.

17

18 On the other hand, students choosing to live alone appear to be relatively older than those  
19 living with roommates. The age variable is highly significant in their living arrangement  
20 and residence distance location models. The distance willing to live from the university  
21 model suggests that the older these student are, the more likely they will live further from  
22 campus. Notably, income of these students is also low. The low household income level  
23 could be attributed to the fact that these are only one-person households comprised of  
24 students.

25

26 Similar to the students who live with roommates, students who live alone are likely to own  
27 a driving license but are unlikely to have personal vehicles. Interestingly, the results  
28 suggest that if they own a vehicle they are more inclined to live closer to campus. Initially  
29 this statement may seem counterintuitive, however, it could hold some truth. For example,  
30 ownership of a vehicle can serve as a proxy for a student's financial status. Provided this is  
31 true, it could be hypothesized that those who own a vehicle are of a higher income status  
32 and thus can afford to live alone and live closer to campus. Neighborhoods in the vicinity of  
33 universities, particularly urban universities, have shown to have higher housing prices  
34 (Cortes 2004) and Toronto is no exception.

35

36 In terms of students choosing to stay in their parental/family home, their decision is largely  
37 influenced by the number of years they have lived in that location and their family's  
38 ownership of the home. It could be postulated that families of those who own the home and  
39 have lived there for a long time have high family and/or home attachment.

**Table2: Empirical Model**

<b>Dwelling Configuration Choice Model</b>									
Variable	Living with parents		Living with a partner		Living alone		living with roommate(s)		
	Para	t-stat	Para	t-stat	Para	t-stat	Para	t-stat	Para
Alternative Specific Constant	0.000	---	-14.399	-17.34	-12.286	-12.29	-7.016	-6.93	
Number of years living in current location (log transformed)	1.026	17.71							
Family owning the home	0.843	9.03							
Household income less than \$30,000 per year	0.000	---	0.000	---	0.506	4.70	0.382	4.18	
Number of members in the household	1.065	17.60	0.648	10.20	0.000	---	0.886	15.40	
Numer of dependent children in the household	-0.060	-0.68	-0.305	-3.10	0.000	---	-0.650	-5.42	
Number of seniors (75+) in the household			0.562	4.04					
Number of cars per adult household members	0.000	---	-0.073	-0.48	-0.246	-1.59	-0.334	-2.17	
Gender: Female			0.233	2.35					
Age in years (log transformed)	0.000	---	4.874	19.67	4.536	15.17	2.299	7.29	
Student: Graduate/Professional			0.335	4.33					
Status: not full time							-0.318	-2.14	
Having a driving lincese	0.000	---	0.417	3.64	0.273	2.16	0.338	2.87	
Expected Maximum Utility of mode choice	0.000	---	0.520	8.93	0.636	8.06	0.957	12.40	

<b>Logarithm of the Distance Willing to Love from the University</b>										
	Living with parents		Living with a partner		Living alone		living with roommate(s)			
	Para	t-stat	Para	t-stat	Para	t-stat	Para	t-stat	Para	
Constant			-1.369	-2.58	-0.991	-1.59	3.267	4.47		
Gender: Female			0.103	1.42						
Age in years (log transformed)			0.883	5.61	0.804	4.32	-0.593	-2.57		
Owning a car			0.105	1.49	-0.130	-2.05	0.000	---		
Expected Maximum Utility of mode choice for considering the following factor as the 'Primary factor'in choosing residence										
-Proximity fo public transit			-0.578	-8.63	-0.654	-11.46	-0.447	-4.38		
-Ability to walk or bike to campus			-0.789	-15.42	-0.802	-14.58	-0.621	-12.57		
-Cost of housing			-0.615	-15.01	-0.653	-11.17	-0.491	-10.09		
-Neighbourhood and proximity of family/friends			-0.595	-13.68	-0.697	-13.06	-0.529	-9.69		
<b>Mode Choice Model of Regular Trips from Home to the University</b>										
	Driving car		Car passenger/pool		Public transit		Bicycle		Walk	
	Para	t-stat	Para	t-stat	Para	t-stat	Para	t-stat	Para	t-stat
Constant	2.373	36.07	0.000	---	1.533	26.44	-0.120	-1.79	-2.282	-22.12
Time	-0.022	-21.31	-0.022	-21.31	-0.022	-21.31	-0.022	-21.31	-0.022	-21.31
Cost/fare	-0.330	-18.87	-0.330	-18.87	-0.330	-18.87				
Loglikelihood of full model -10219										
Loglikelihood of null model -7839.5										
Number of parameters in full model 57										
Number of parameters in null model 4										
Adjusted Rho-Squared value 0.23										
<b>Corelation between unobserved factors affecting resident configuraion and location</b>										
	Para	t-stat								
Correlation coefficient	-0.93	-141.27								

1 The household size and the number of seniors over the age to 75 in the parental/family  
2 home have considerable positive effects on a student's decision to stay home. Intuitively,  
3 this makes sense. Larger families could impose greater responsibilities and chores on the  
4 student. Also, the student may need to assist and care for the senior family members.  
5 Interestingly, this finding substantiates the statistic mentioned earlier in the paper that  
6 17% of survey respondents said their residence location choice was beyond their control.  
7 The number of dependent children in the household has a negative influence on a student's  
8 probability of living with parents/family. One plausible rationale could be that students are  
9 more inclined to leave their family home because the presence of children makes a loud,  
10 distractive environment, not ideal for studying.

11  
12 Students deciding to live with a partner typically comprise of older students and students  
13 with a graduate or professional student status. Similar to students who live alone, the  
14 group's willingness to live further from campus increases with age. The distance willing to  
15 live from the university model also reveals that those in the group who own a car are  
16 willing to live even further from campus. However, the number of cars per adult household  
17 member variable exerts a negative, though small influence on the student's probability of  
18 living with a partner. Furthermore, driving license ownership exhibits a positive effect on  
19 the living with a partner choice option. This could suggest vehicles are likely shared among  
20 partners. Interestingly, the group's probability of living further from campus also increases  
21 if the student is a female. Moreover, the living arrangement choice model shows a  
22 statistically significant positive correlation between female students and the choice of  
23 living with a partner. The number of members in the household appears to have a strong  
24 explanatory power in the living with a partner configuration model. However, the number  
25 of dependent children in the household is negatively correlated with the probability of a  
26 student living with a partner. It is possible that those living with their partner and  
27 attending school are unlikely to have children at that time.

28  
29 Across all living arrangements analyzed in the residence location choice model, it is  
30 apparent that student's expected maximum utility of mode choice has a profound negative  
31 effect on the distance students are willing to live away from the campus. The walk and  
32 cycling accessibility of a location appear to be the most influential factor in student's  
33 residence location choices. However, other factors such as the cost of housing,  
34 neighborhood environment and proximity to friends/family, and proximity to transit  
35 follow closely behind.

36  
37 In the context of policy relevance, we have achieved a better understanding of the  
38 motivations and dynamics at play in student's residence choices. The findings of the study  
39 suggest that to promote active mode commutes, policy initiatives should be in place to  
40 ensure affordable housing options in the proximity to campuses. One method to achieve  
41 affordable housing is to increase the residential density permitted in the area as well as

1 take advantage of the housing's occupancy standards. It is clear from the results that  
2 students, particularly full-time undergraduate students, are willing to compromise living  
3 space for convenient commutes to campus. Therefore, there is demands for housing with  
4 multiple shared dwelling units around the university areas. As for graduate/professional  
5 students who tend to live with their partners or live alone, housing options for single and  
6 two-person households should also be provided near the university. However, those living  
7 alone or with their partners appear less sensitive to commuting distance compared to  
8 those living with roommates, thus there should be provisions to ensure a larger percentage  
9 of high density, shared housing options adjacent to the university.

10  
11 However, the land use of the area should not be entirely dominated by residential housing  
12 for students. As shown in the study, the neighbourhood environment is an important factor  
13 that students consider when choose their residence location. Thus, mixed use is  
14 encouraged and it also improves the walkability of the area. Furthermore, planners should  
15 ensure the neighbourhoods around the university are highly pedestrians and cyclists  
16 friendly as it appears students are inclined to choose these active modes of transport when  
17 they are perceived as feasible options. Transit accessibility is important as well for the  
18 majority of university students in Toronto use transit to commute.

19  
20 The study does show that many students hold a driving license though they do not typically  
21 own a vehicle, thus there may be a market for a car-share/car-rental program around  
22 university campuses. In terms of the students who do not have a choice but to drive to and  
23 from campus, universities could consider implementing carpool programs to reduce the  
24 parking demand on campus. When creating these carpool programs, the factors motivating  
25 current student carpoolers that are revealed in the study should be considered: save on  
26 parking cost and time, and proximity to carpool partner(s).

## 27 28 **6. Conclusions and Recommendations for Future Research**

29  
30 This paper investigated the typical commuting mode choice, living arrangement choice, and  
31 residence distance location choice of students from four universities in Toronto. An  
32 econometric model is proposed that jointly models these three choices. The overall  
33 objective of the study is to identify the nature and extent of various systematic and random  
34 factors influencing university student's residence choice behaviour in order to provide  
35 evidence for the planners and policy makers to act upon. Clearly, the implicit and  
36 apparently un-captured (by systematic variables) trend is that students who tend to leave  
37 the family home to attend the university would prefer to live close to their campus;  
38 however, the urban transportation and land use contexts do not allow them to do so  
39 optimally. As a result, the trade-off between choices of living configuration and the distance  
40 they are willing to live from the University are influenced by these transportation, land use  
41 and socioeconomic factors.

1  
2 The resulting model revealed modal accessibility significantly influence the living  
3 arrangement and the residence distance location models. It was found that students  
4 choosing to live alone or with roommates are likely to live in low-income households which  
5 often have a lower vehicle to household member ratio. Students living with roommates  
6 appear to be highly sensitive to modal accessibility. They are also more likely to be younger  
7 and hold a full-time student status. On the contrary, those living alone or with a partner  
8 tend to be older students, possibly with a graduate or professional student status. The  
9 probability of students choosing to stay at home with their parents/family is positively  
10 influenced by the size of the household and the number of seniors (over the age of 75) in  
11 the household.

12  
13 For students who are more probable to leave their parental/family home, a residence  
14 distance location choice model was developed to estimate the distance they are willing to  
15 live away from campus. It was found that for those living with roommates, the younger  
16 they are the more inclined they are to live closer to campus. However, for students living  
17 with partners and living alone, their likelihood of living closer to campus decreases with  
18 age. Intuitively, owning a car allows those living with a partner to be less sensitive to  
19 distance in their location choice.

20  
21 Across all living arrangements, it was apparent that student's expected maximum utility of  
22 mode choice has a profound negative effect on the distance students are willing to live  
23 away from campus. The walk and cycling accessibility of a location appear to be the most  
24 influential factor. Additional factors such as the cost of housing, neighborhood environment  
25 and proximity to friends/family, and proximity to transit also have significant effects. This  
26 suggests that affordable housing choices by the university campuses can significantly  
27 increase the number of active mode users among the student population in Toronto.

28  
29 This study contributes to a better understanding of university students' residence choice  
30 and commuting behaviour. However, the framework in this paper can be extended to  
31 estimate student's discrete residence choices, as opposed to using a continuous distance  
32 variable. This could help provide more insight on the spatial distribution of students'  
33 housing choices and the impact of housing characteristics, such as rental cost, on these  
34 choices.

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