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# 1 Urban Data Science for Sustainable Transport 2 Policies in Emerging Economies

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## 22 — **Abstract** —

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23 In the city of Hanoi, Vietnam, as with other rapidly-developing cities, transport infrastructure is  
24 failing to keep pace with the burgeoning population. This has led to high levels of congestion,  
25 air pollution, and a broad inequity in the accessibility of large parts of the city to residents. The  
26 emerging discipline of Urban Data Science has a valuable role in providing policy makers with robust  
27 evidence on which to base policy, but the discipline faces problems with the application of techniques  
28 that are based on assumptions that do not hold when applied to emerging economies.

29 This paper presents the preliminary outputs of a new programme of urban data science work  
30 that is being developed specifically for Hanoi. It leverages a spatial microsimulation approach to  
31 up-sample a bespoke travel survey and create a synthetic representation of the transport preferences  
32 of all residents in the city. These new data are used to assess the impacts that changes in the broader  
33 socio-economic context, such as increasing prosperity amongst residents, could have on rates of car  
34 ownership and hence on the problems of congestion and pollution. The results begin to highlight  
35 parts of the city where the impacts of improved economic conditions coupled with changes to wider  
36 transport policies might lead to greater use of personal cars in the future.

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## 43 **1** Introduction

44 The structure of cities and transport systems are closely related and road networks play a  
45 key role in meeting the transport needs of urban areas. However, Hanoi, Vietnam, like many  
46 major cities in emerging economies, suffers serious traffic congestion and air pollution due to  
47 rapid urbanization, increases in private transport, and the informal infrastructures formed  
48 during the emergence of urban sprawl. The field of Urban Data Science (UDS) consists of  
49 “quantitative workflows for gathering, processing, and analyzing data in a spatiotemporal  
50 context that applies statistics and computer science to urban questions” [2], and could be  
51 extremely valuable as a means of better understanding the dynamics of cities such as Hanoi.  
52 Hence the application of UDS to questions about sustainable transport infrastructure could  
53 help to generate robust, effective policy to reduce the burdens of traffic congestion and  
54 pollution.

55 However, there are some fundamental difficulties with the conceptualisation of UDS that  
56 present challenges for the application of core UDS techniques that have emerged in the  
57 Global North to cities in emerging economies [1], such as Hanoi. For example, data that  
58 are commonly used in the application of UDS to urban mobility include those that are  
59 created from the use of smart cards and intelligent transportation systems [2], i.e. “analytics-  
60 powered, intelligent traffic management” [8]. But this does not translate to systems where  
61 the commonly-used means of transport either do not record information about journeys  
62 digitally (e.g. via cash-based ticketing systems) or are fundamentally organised in much  
63 more ad hoc manner that lack any formal means of recording journeys or even publishing  
64 timetables (such as the matatu system in Nairobi [15]). Similarly, commonly-used transport  
65 modelling methods, including both aggregate traffic assignment [12] and microsimulation  
66 models [9], will struggle to account for the behaviour of vehicles such as motorbikes that  
67 do not follow the behaviours that would be expected from car drivers but are extremely  
68 common in many emerging economies.

69 This paper reports on the work as part of a wider project that aims to use UDS techniques  
70 to provide policy makers at the highest level of government with new data and computer  
71 models to support evidence-based policy to create a more efficient, equitable, and sustainable  
72 transport system that meets Hanoi’s expanding population needs. In the context of limited  
73 data availability with regards to the dynamics of the transport infrastructure – particularly  
74 when compared to cities of a similar size that are characteristic of those in the Global North  
75 – we fall back on the creation of a *synthetic* population that is designed to represent all  
76 individuals in the city of Hanoi. The population is created through the use of simulated  
77 annealing as a means of up-sampling a new survey of 1,500 households (conducted specifically  
78 for the project) by combining it with data from the 2019 Vietnam population census. Initial  
79 results begin to provide an insight into the preferences of residents for different types of motor  
80 vehicle use, highlighting areas that are at particular risk of becoming more car dependent as  
81 households become more affluent, or as the nature of the transport infrastructure changes.  
82 More broadly, the project aims to explore the relevance of commonly-used UDS techniques  
83 in the context of a rapidly developing city in an emerging economy.

## 84 **2** Research Context

85 In Hanoi, motorbikes are the preferred transportation mode: over 90% of the vehicles driven  
86 in Hanoi are motorbikes and there are on average 2.5 motorbikes per person [14]. Since  
87 the introduction of the Doi-Moi policy [6] in the 1980s, the number of motorbikes has  
88 increased 10-fold and there are now more than 4 million motorbikes in Hanoi alone [5, 6].

89 Simultaneously, public transport infrastructure has developed slowly. As public transport  
90 does not meet the city’s requirements, increases in personal traffic are inevitable, resulting in  
91 acute welfare problems, especially air quality. Pollution is chronic, with PM2.5 and ozone  
92 concentration regularly exceeding safe levels. In response, the City has developed fast buses,  
93 a skytrain system, tightened the standards for new vehicles and imposed petrol quality  
94 controls. Nevertheless, the Real-time Air Quality Index, measured by the U.S. embassy,  
95 recently found pollution in Hanoi at levels sufficient for people with heart and respiratory  
96 problems to stay indoors. Some officials proposed a radical plan to ban motorbikes in large  
97 parts of the city, but this was met with strong public opposition. Surveys linked to models  
98 can answer questions about how, where, and when motorbikes should be banned (if at all),  
99 about the impacts on local communities, whether public transport can cope, and whether  
100 there are better alternatives. Importantly, they can also provide information about the  
101 factors that are encouraging or prohibiting peoples’ vehicle ownership preferences; this paper  
102 pays particular attention to the factors that might lead to greater car ownership amongst  
103 residents, especially if motorbikes are no longer an option for travel.

## 104 **3 Data & Methods**

105 The aim of this work is to up-scale a new survey of transport behaviours and preferences  
106 conducted by the project team in Hanoi, Vietnam. We do this through linkage to the  
107 most recent Vietnamese population census using simulated annealing to create a synthetic  
108 population of all individuals in Hanoi that contains core census variables as well as variables  
109 in the new survey. The following sections outline the methods used; drawbacks and caveats  
110 are discussed in Section 5.

### 111 **3.1 Transport Survey**

112 A new survey is currently being conducted in Hanoi that asks people for basic demographic  
113 information as well as details about their travel behaviour (e.g. common journeys) and  
114 preferences (e.g. aspirations for ownership of different types of vehicle). The COVID-19  
115 pandemic has interrupted the survey on multiple occasions, but at the time of writing 1,500  
116 households, out of a target of 10,000, have responded. The key variables that are relevant  
117 for this paper include, among others:

118 **Demographics** Sex, age group, occupation.

119 **Vehicle ownership** Types of vehicles owned by the household.

120 **Travel behaviour** Details about regular journeys made: start/end locations, frequency, mode  
121 of transport, reasons for choice of mode.

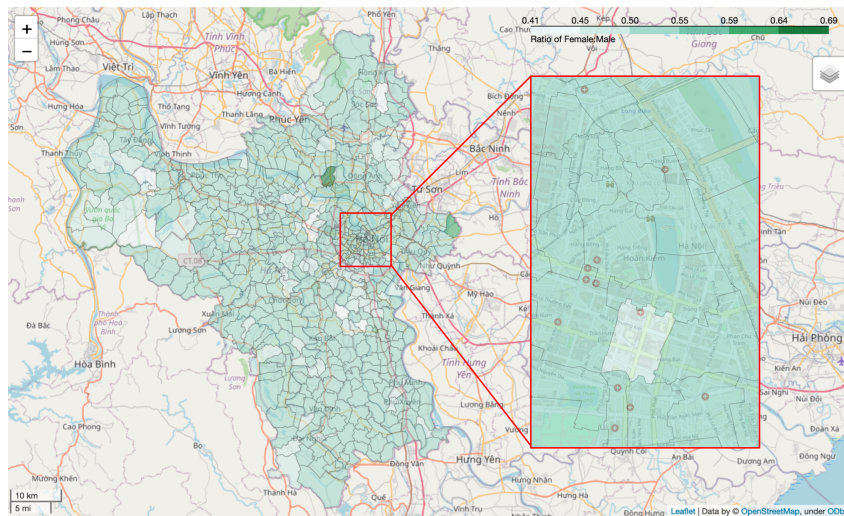
122 **Vehicle ownership aspirations** Whether the household would like to own additional vehicles  
123 and what factors prevent them from ownership.

### 124 **3.2 Vietnamese Census of Population**

125 Vietnam’s most recent population and housing census was conducted in 2019 [4]. It found  
126 that the population of Vietnam had grown to 96M people. Hanoi, the case study area, is  
127 the second largest city after Ho Chi Minh City with a population of 8M people; increasing  
128 by 1.5M between 2010 and 2019<sup>2</sup>. At the time of writing, the project has access to counts

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<sup>2</sup> <https://vietnam.opendevelopmentmekong.net/topics/vietnams-population-and-census/>



■ **Figure 1** The ratio of females to males in Hanoi from the 2019 Vietnam census.

129 of people per district level broken down by sex and age group (as separate variables, not  
 130 cross-tabulations). As an example to demonstrate the level of geography available in the  
 131 census, Figure 1 illustrates the ratio of males to females in each district in the study area.

### 132 3.3 Synthetic Population Generation

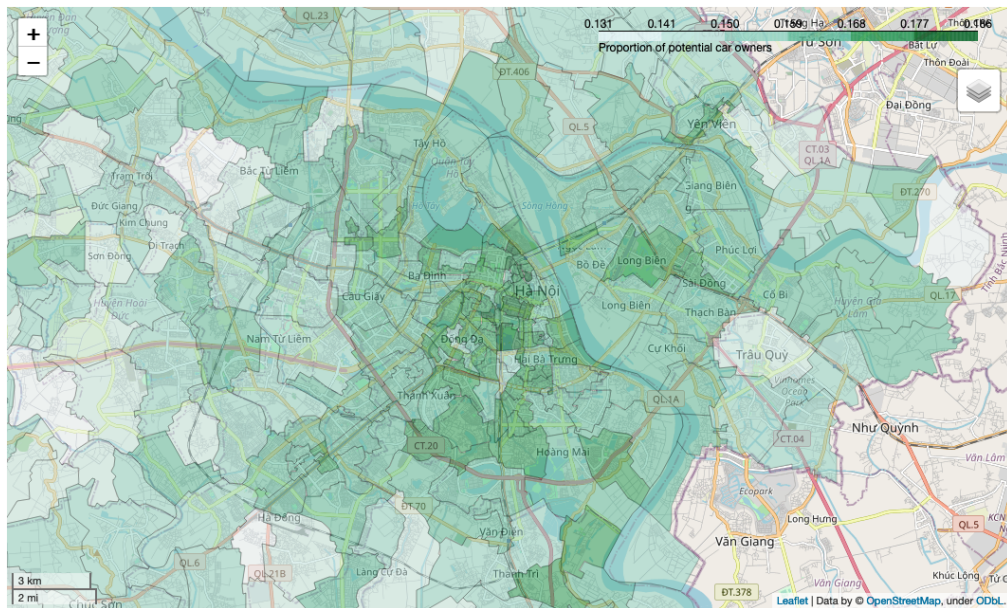
133 The survey (Section 3.1) aims to include responses from 10,000 households which will be  
 134 one of the largest household travel surveys conducted in Hanoi. However, the city is so  
 135 populous that, naturally, the geography of the respondents is very sparse (there will be very  
 136 few respondents per district). Therefore to up-scale the survey in order to make inferences  
 137 about transport behaviour across a much wider spatial area, we use population synthesis  
 138 to combine the survey results with the population census to create a synthetic population.  
 139 The new population aims to be representative, both in terms of demographics and transport  
 140 behaviours, of the true underlying population.

141 Synthetic population generation (also sometimes referred to in geography as ‘microsimu-  
 142 lation’) is an alternative to ‘zonal’ disaggregation methods [3] that was originally inspired by  
 143 the work of Orcutt [13]. It aims to construct a data set of individual units (people in this  
 144 case) over a large area by cloning individuals from a survey (the travel survey) such that  
 145 the aggregates match some known aggregate data (the population census). The resulting  
 146 synthetic population contains attributes from both the aggregate and survey data [7]. The  
 147 assignment of individuals to areas is conducted using an iterative optimisation algorithm  
 148 simplified from simulated annealing [10] as implemented in the Flexible Modelling Framework  
 149 software<sup>3</sup> [7]. The work will not currently attempt to assign individuals to specific buildings.

## 150 4 Results

151 The survey is extremely rich, so there are a wide variety of variables that are attached to the  
 152 synthetic population and could be analysed. Here we examine one factor; the propensity

<sup>3</sup> <https://github.com/MassAtLeeds/FMF>



■ **Figure 2** The proportion of people in the synthetic population who would like to own a car but are prohibited from purchasing one due to the cost.

153 for individuals, who do not currently own a car, to purchase one. Individuals are extracted  
 154 from the population who meet three criteria: (i) they do not currently own a car; (ii) they  
 155 would like to own a car in the future; (iii) cost is the main factor that is prohibiting them  
 156 from owning one. Figure 2 then maps the proportion of synthetic individuals who meet  
 157 the criteria above. Although these are preliminary results and it is too soon to draw firm  
 158 conclusions, it is interesting that towards the centre of the city there are larger proportions  
 159 of synthetic individuals who would like to purchase a car if they could afford to. This issue is  
 160 important for policy because, as the economy in Vietnam expands and more people become  
 161 able to afford a motor car, transport policies will need to encourage alternative means of  
 162 transport to prevent an unsustainable rise in car use.

## 163 **5 Discussion & Conclusions**

164 The field of Urban Data Science (UDS) has shown promise as a means of better understanding  
 165 the dynamics of cities in order to make them better places to live. However, assumptions  
 166 about data characteristics and availability do not necessarily translate well to the urban  
 167 context in developing economies. In Hanoi, for example, there are very limited digital data  
 168 that describe the use of the transport network. Therefore this paper leverages a synthetic  
 169 population generation framework to up-scale a new transport survey, allowing inference  
 170 about transport behaviours over a much wider spatial area than would be possible otherwise.  
 171 Preliminary results suggest that the distribution of residents who have the *propensity* to own  
 172 a motor car (i.e. they would own one if they could) varies considerably across the city. This  
 173 has the potential to inform transport policy, providing robust data to support sustainable  
 174 transport policy.

175 This is preliminary work and there are many caveats that need to be resolved. To begin  
 176 with, the survey needs to be distributed to a wider population in Hanoi. Secondly, there is  
 177 a discrepancy between *households* and *individuals*. The survey collects information about



178 households, but currently the synthetic population generation algorithm creates synthetic  
 179 individuals, not households. Future work aims to take an additional step that will allow the  
 180 synthetic individuals to be grouped into households, following [11]. Thirdly, the currently  
 181 available census data contain only counts of people by age group. Hence age group is the  
 182 only constraint used in the creation of the synthetic population, which means that the work  
 183 assumes that all people in a particular age group will have similar behaviours and preferences  
 184 with respect to transport use. This is obviously a very weak assumption. To make the  
 185 analysis more robust, future work will make use of census data that contain a much richer  
 186 set of cross-tabulated variables, as well as additional variables that are present in both the  
 187 survey and the census that can be held back from the synthetic population generation process  
 188 and used as a means of validation.

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