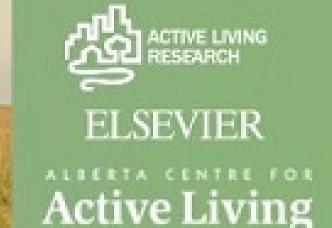
Create change









# A novel approach to investigate the impact of the built environment on physical activity among young adults

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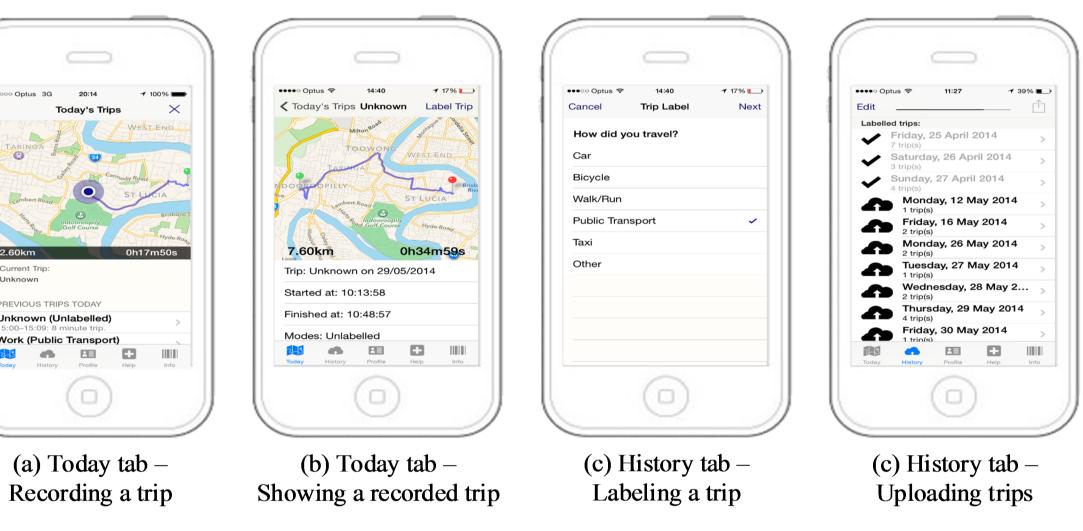
#### Introduction

- The **built environment** can "facilitate" or "hinder" **physical activity (PA)**, including walking and cycling for transport purposes.
- The greatest majority of the evidence relies on **self-reported measures** of PA.
- **Questionnaires** have a **low level of accuracy** and are more prone to **self-reporting biases**, compared to devices such as accelerometers and pedometers which serve to objectively measure PA.
- Relying on the technological advancement and widespread accessibility of smartphone, this study introduces a **custom-designed smartphone app** to **collect PA data during transport** and, shows how the data collected by the app can provide us with **insights** about young people's **transport-related PA** in relation to the built environment and trip characteristics.
- Understanding PA during travel is relevant to support investments and programs that **support sustainable modes of transport**, such as walking and cycling.

#### Methods

**Data Collection:** Designed and implemented a smartphone app for both iOS and Android platforms:

- automatically captures all movements of its users;
- includes post-processing algorithms that detect the types of activities within a trip



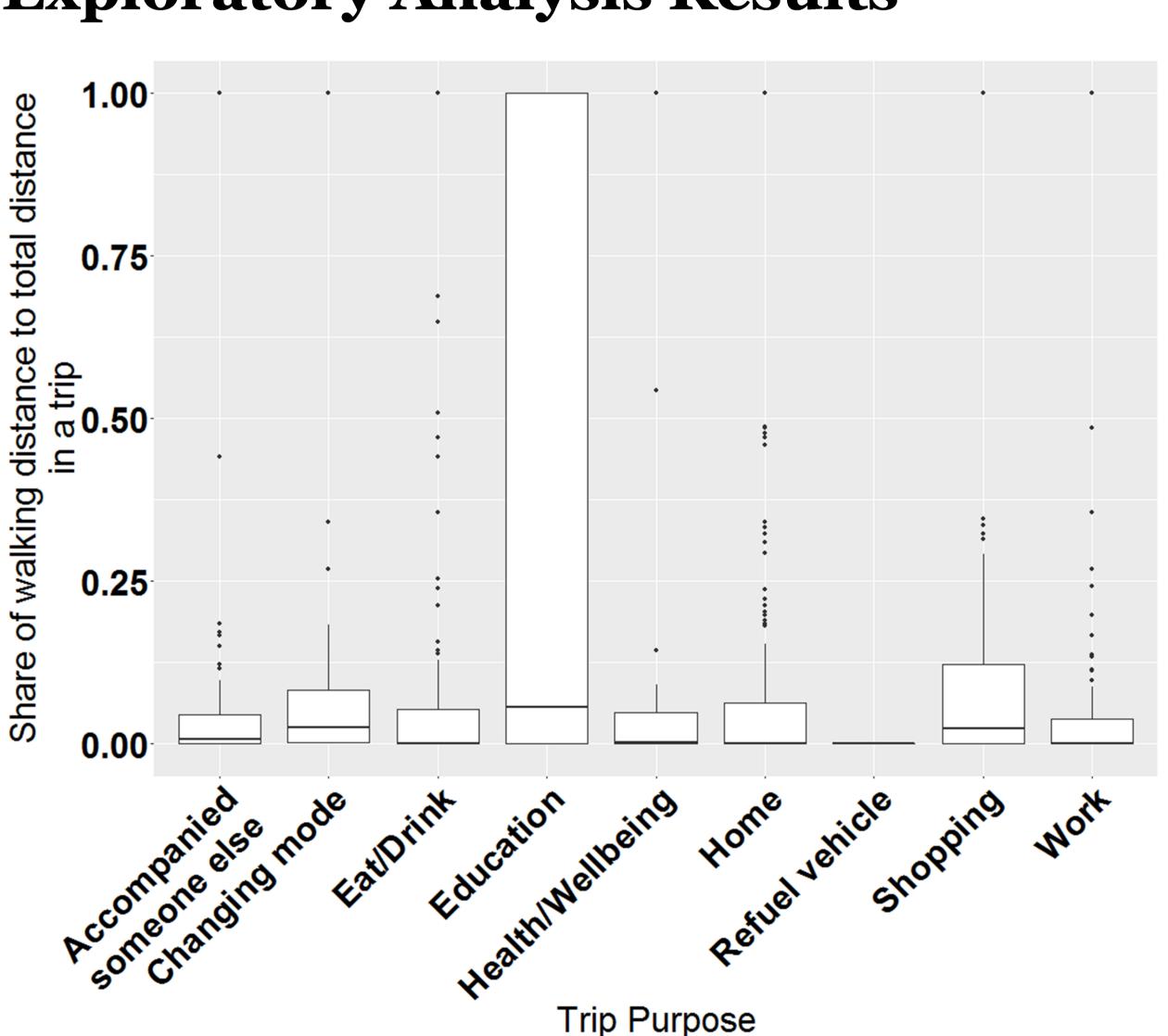
Data: Data from 142 university students in Brisbane, Australia:

- collected for an average of 3 days per participant;
- include 422 person-day observations and a total of 2024 single modal activities

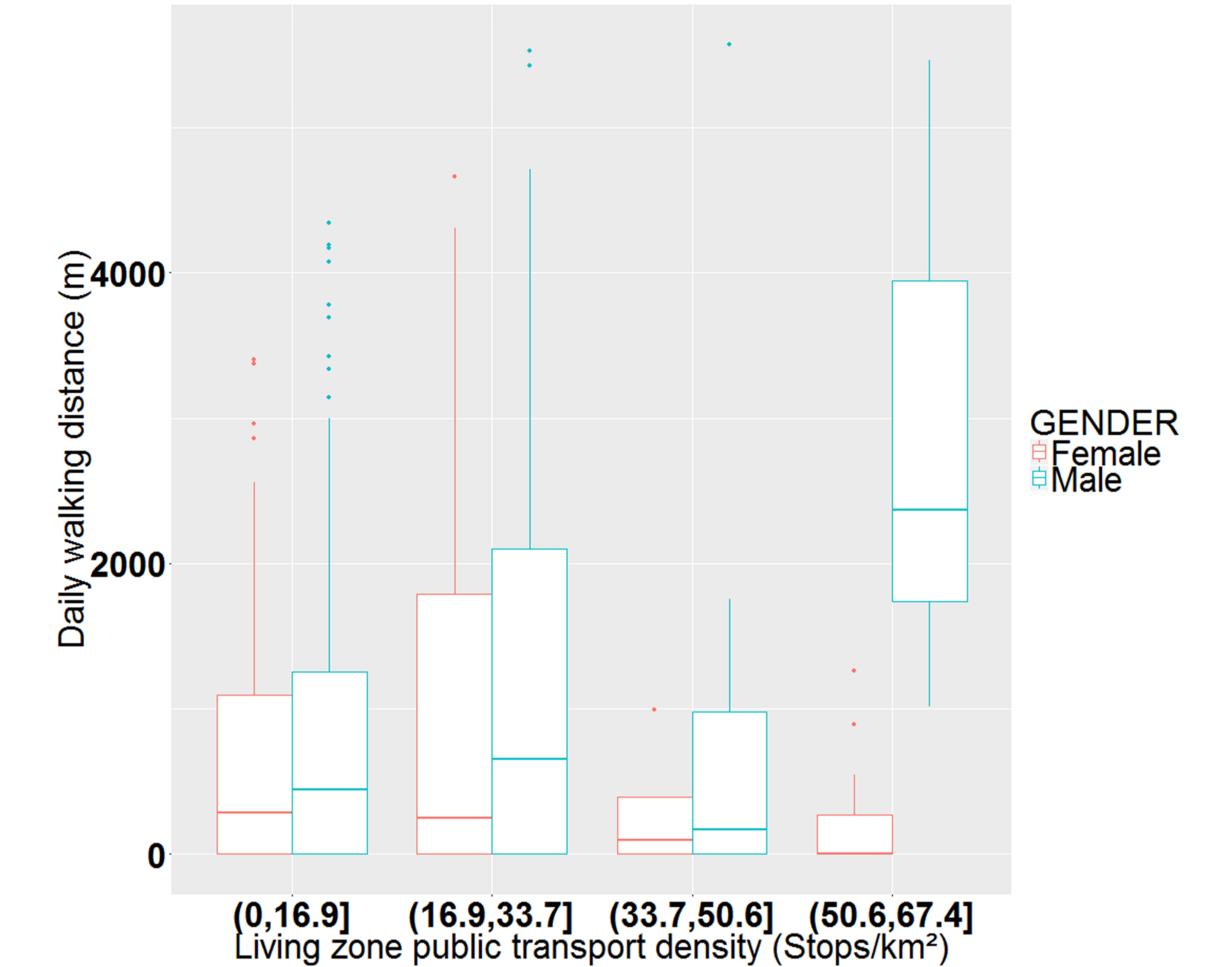
**Data Analysis:** We conducted statistical and spatial analysis using R, RStudio and QGIS including:

- descriptive analysis and exploratory analysis;
- pooled ordinary least squares analysis with panel data to model the impact of built environment attributes and trip characteristics on transport-related PA.

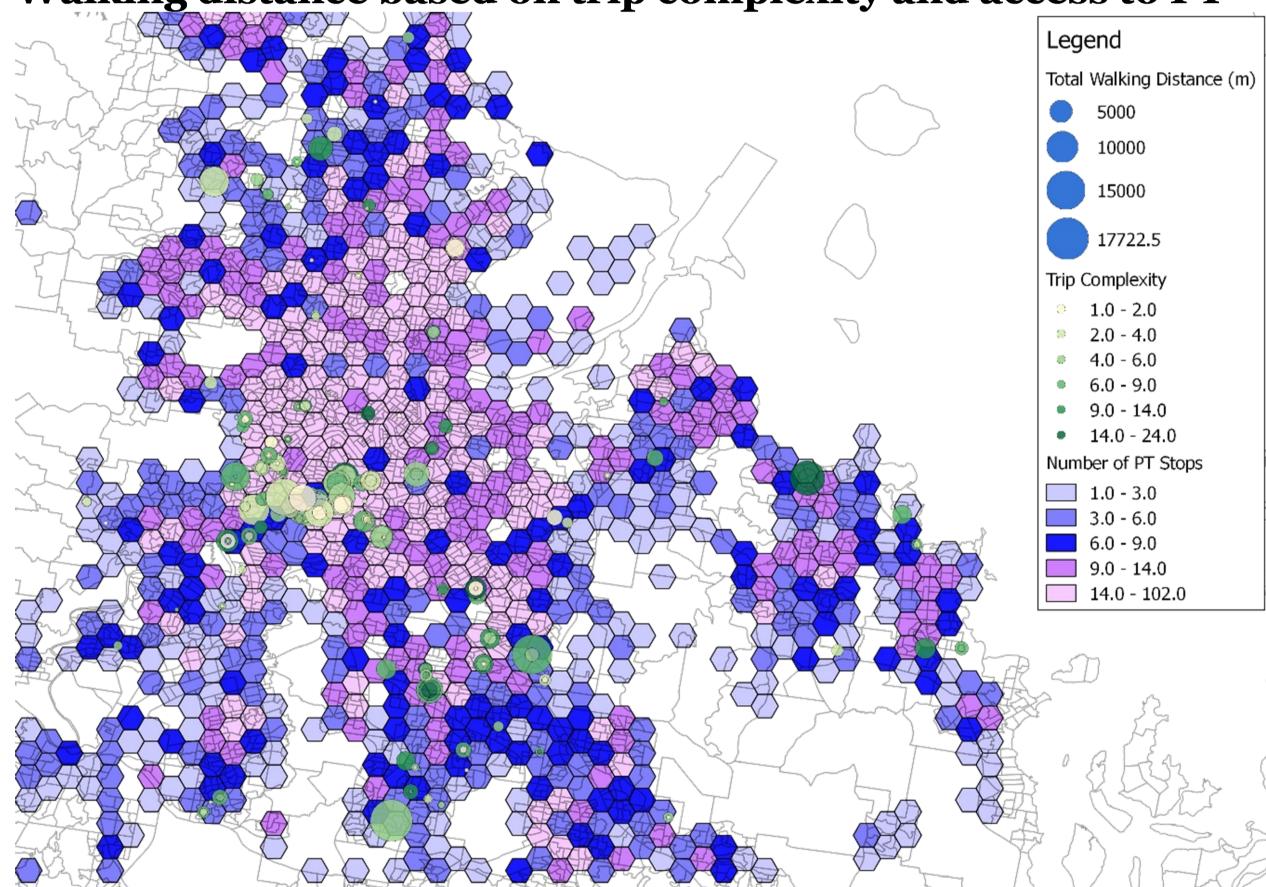
# **Exploratory Analysis Results**



#### Walking distance, public transport stop density and gender



Walking distance based on trip complexity and access to PT



## **Pooled OLS Model**

**Dependent Variable: Walking Distance** 

			Std.	t
Variable		Estimate	Error	– value
Intercept		1296.545*	736.403	1.761
Home public transport accessibility		0.078	0.631	0.123
(mean distance to 5 closest stops $-m$ )				
Stop Density	$16.9 < Density <= 33.7 / km^2$	59.027	283.462	0.208
	$33.7 < Density <= 50.6 / km^2$	1276.07**	568.826	2.243
	$50.6 < Density <= 67.4 / km^2$	326.373	554.076	0.589
Trip complexity (triplegs / travel day)		231.2****	43.85	5.272
Total distance travelled per day (m)		-0.012***	0.004	-2.665
Car access		-1040.154****	278.603	-3.733
Bike access		222.604	387.736	0.574
Age	21-30 years	-379.138	256.084	-1.481
	41-50 years	-2164.701***	662.563	-3.267
	51 years or more	-1101.594	928.03	-1.187
Job – Working		1606.517****	449.64	3.573
Weekly Income	\$200-\$299	-89.353	294.111	-0.304
	\$300 or more	-667.665**	316.699	-2.108
HH Siz	2 – 4 persons	-102.075	515.587	-0.198
	5 or more persons	-336.513	528.579	-0.637
Gender – Male		56.011	260.36	0.215
SA1 per cent population in medium dis/advantaged quantile (40%-60%)		14.891*	7.764	1.918
Table	description: * $p < 0.1$ , ** $p < 0.05$ ,	*** $p < 0.01$ , **** $p$	< 0.001	

 $R^2 = 0.215$ , Adjusted  $R^2 = 0.165$ , F(18,282) = 4.288, p < 0.001 Max VIF = 3.20