Estimating a Toronto Pedestrian Route Choice Model using Smartphone GPS Data

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Presentation Outline

- Introduction
- Background
- Data
- Smartphone Data
- Alternative Route Generation
- Choice Model
- Toronto Case Study
- Results
- Route Generation Analysis
- Conclusions

1. Introduction

Study Motivations

- Travel demand models overlook walking trip routes
- City planning supports building walkable streets but measures are often qualitative
- Smartphone GPS surveys are becoming more common for data collection

2. Background

Built Environment

- Built environment Buildings, transportation systems, open space, and land-use that support communities and impact human health (City of Toronto, 2015)
- Various measures:
	- Perceived measures
	- Observed measures
	- Geographic measures

Built Environment and Pedestrian **Travel**

- Effects of built environment on walking rates
- Effects of built environment on walking routes
	- Very few studies
	- Mainly qualitative

Built Environment and Pedestrian **Travel**

- Guo (2009)
	- One more intersection per 100m increased utility by 0.3 min, increasing sidewalks by 6ft increases utility by 0.5 min, and people willing to walk 2.9 min to avoid hilly topography
- Dill and Broach (2015)
	- turns equivalent to +50m, upslopes of 10% are twice as costly, unsignalized arterial path perceived as +70m, busy roads 14% longer, commercial neighborhoods 28% shorter

3. Data

Street Network Data

- Toronto Open Data
	- Street Network
	- Sidewalk Conditions
	- Signalized Intersection Locations
	- Land Use
- Elevation
- Walk Score

Walk Score

• Considers proximity to amenities, walking infrastructure, population density, block length, intersection density

Walk Score

Land Use

- Address point with land use
- Land parcel

Need to merge these files and convert into a "land use frontage" measure

Land Use Comparison

Land Use

4. Smartphone Data

Smartphone Data

- Collected during the Waterfront Project in 2014
- 4 week survey period starting in November
- Passive GPS location
	- Records location after 50m of travel distance from previous point

Smartphone Data

- Post Survey Data Processing
	- Trip ends determined based on 3 minute dwell time
	- Travel modes were inferred based on speed profiles (87% success rate for mode detection)
	- Trip purpose was not collected

*Outlined in paper by Harding, Zhang, & Miller (2015)

• 3193 walking trips across 103 individuals

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- Remove trips with large gaps (200m)
- Remove trips with 3 or less points
- Remove mislabelled walk trips

Large Gap Trips

- Check gaps if they coincide with subway stations
- Break trip into two walking trips

Walk Trip Solving Process

Walk Trip Solving Process

(Dalumpines & Scott, 2011)

Map-Matching Issues

- Pedestrian trips can go through buildings or open spaces
- Alternate routes may exist within buffer area
- Large gaps may make buffer area not continuous
- Filling GPS points in straight line may cut corners

Walk Trip Issues

Walk Trip Issues

4. Alternative Route Generation

Stochastic Route Generation

- Biased random walk algorithm
- Builds the route link by link, making its way to the destination
- At each node it assesses the next links to take
- Probabilities of each branching link are determined
- Monte Carlo simulation decides which link is chosen

1. Import origin and destination

2. Determine origin street segment

3. Find the street segments connected to the source node

4. Determine the cost for each street segment

5. Determine the cost from the source node to the destination

6. Calculated probabilities and use Monte Carlo simulation to select next segment

7. Repeat process for newly selected segment and source node

8. Once destination segment is reached, stop process and generate route

Route Generation Rules

$$
P(i) = \frac{1 - \left(1 - \left(\frac{SP(v, D)}{cost(i) + SP(w, D)}\right)^{\alpha}\right)^{\beta}}{\sum_{i \in M} 1 - \left(1 - \left(\frac{SP(v, D)}{cost(i) + SP(w, D)}\right)^{\alpha}\right)^{\beta}}
$$

Where:

Probability of choosing link i out of possible outgoing links (M)

Source node v and sink node w

SP(v,D) is the shortest path/least cost path from source node v to destination D

Cost(i) is the cost of link i

α and β are parameters that make the probability more sensitive to increase in cost.

(Freijinger, 2007)

Route Generation Rules

- No node is traversed twice. If a loop is detected, the route generation attempt fails.
- U-turns are not needed
- The generated path does not exceed two times the shortest path between O and D
- The route does not pass the destination link
- If a dead end is reached, the route generation attempt fails and the dead end segment is recorded so it is not considered again. After 10 attempts, the iteration is abandoned
- Travel on street segments that go in a direction away from the destination are heavily penalized (cost=9999m) unless they are on the shortest path from the source to the destination.

Route Generation Rules

- Additional Modifications
	- Turns equivalent to +50m
	- Travel on streets with complete sidewalks is 10% shorter

5. Choice Model

Path Size Logit Model

$$
P(i|C_n) = \frac{e^{\mu(V_{in} + \ln(PS_{in})) + \ln\left(\frac{k_{in}}{q(i)}\right)}}{\sum_{j \in C_n} e^{\mu(V_{jn} + \ln(PS_{jn})) + \ln\left(\frac{k_{in}}{q(j)}\right)}}
$$

Where:

Cn is the choice set for user n (includes chosen route)

μ is the logit scale term

 V_{in} is systematic utility for alternative i for user n

 PS_{in} is the expanded path size factor for alternative i for user n

 k_{in} is the number of times alternative i is randomly drawn. If chosen route, $k_{in}+1$

q(i) is the probability of choosing a route containing the street segments. It is calculated as the product of each link choice probability

(Freijinger, 2007; Frejinger, Bierlaire, and Ben-Akiva, 2009)

Path Size Logit Model

$$
PS_{in} = \sum_{a \in \Gamma_i} \frac{L_a}{L_i} \frac{1}{\sum_{j \in C_n} \left(\frac{L_i}{L_j}\right)^{\Phi} \delta_{aj}}
$$

Where:

Гi is the set of links in path i La is the length of link a Li is the length of path i Lj is the length of path j δ_{ai} equals 1 if link a is on path j and 0 otherwise φ is a parameter that controls the impact of route length in the correction factor

(Ramming, 2002)

6. Toronto Case Study

Route Characteristics

Route Variables

land, high residential land, industrial land, institutional land

Socioeconomic Interaction Terms

- Gender
- Age
- Student Status
- Employment Status
- Income level
- Time of Day

Observed Trips

5. Results

General Model Results

The utility equation for route i is given be the following equation:

 $U_i = \beta_{Length} * Length_i + \beta_{Turns} * Number of Turns_i$ $+ \beta_{Sidewalk}$ Both Sides * Length Sidewalk Both Sides_i $+$ β _{Siginalized} Intersection * Number of Signalized Intersections_i + β_{PS} * $ln(PS_i)$

$$
+ \ln\left(\frac{k(i)}{q(i)}\right)
$$

General Model Results

* all coefficients significant at p<0.05

Attribute

Turn

Signalized Intersection

Sidewalk Both Sides

+33m

Attribute

Turn

Signalized Intersection

Sidewalk Both Sides

-36m

Attribute

Turn

Signalized Intersection

Sidewalk Both Sides

-33% distance

Non-Significant Variables

- Land use
- Development density
- Steep slopes
- Walk Score

Interaction Model Results

The interaction term model's utility equation for route i is given by the following equation:

 $U_i = \beta_{Length} * Length_i + \beta_{Length \text{ Female}} * Length_i * General \text{Gender}_{Female} + \beta_{Turns} * Number \text{ of Turns}_i$ $+ \beta_{\text{Sidewalk Both Sides}} * Length Sidewalk Both Sides_i + \beta_{\text{Signal} instead Intersection}$ $*$ Number of Signalized Intersections_i + $\beta_{Minor \text{ Arterial Student}}$ $*$ Length on Minor Arterial_i $*$ Student + $\beta_{\text{Arterial Age 25}}$ $*$ Length on Arterial_i $* Age$ under $25 + \beta_{Minor$ Arterial Income $*$ Length on Minor Arterial_i $*$ Income over \$75,000 $+ \beta_{\text{Parks} \, \text{Evening}} * \text{Length in } \text{Park}_i * \text{Evening} + \beta_{\text{Commercial} \, \text{Employee}}$ $*$ Length along Commercial Land_i $*$ Employed + $\beta_{\text{Collecter Age 45}} *$ Length on Collector_i $* Age$ over $45 + \beta_{Office \textit{Age 25}} * Length \textit{ along } Office \textit{ Land}_i * Age \textit{under 25}$

> + $\beta_{Walkway \, Age \, 25}$ * Length along Walkways $_i$ * Age under 25 + β_{PS} * $\ln(PS_i)$ + \ln $k(i$ q(i

Interaction Model Results

* all coefficients significant at p<0.05

Table - Average percent difference compared to observed route

Table - Average percent difference compared to observed route

Calibrated route generation method had longer routes, routes with more turns, and more travel on streets with complete sidewalks

- Route generation biased around shortest path had the highest probability of generating observed route
- Calibrated route generation methods were more likely to generate observed route for longer routes

7. Conclusions

Conclusions

- Smartphone GPS data proved to be viable source for pedestrian route choice
- Distance, turns, complete sidewalk, and signalized intersections are significant factors
- Calibrating the stochastic route choice generator made generating the observed route more likely for longer routes

Limitations/Future Work

- GPS accuracy was too low to determine detailed trip behaviour
- Trip purpose not collected
- Stochastic route choice generation works well but may generate very random routes
- Multiple observations influence results
- Land use measure could be improved with observational data

8. References

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