### Activity choice in pedestrian facilities

#### Antonin Danalet

Transport and Mobility Laboratory School of Architecture, Civil and Environmental Engineering Ecole Polytechnique Fédérale de Lausanne

TU Eindhoven visiting EPFL

July 9, 2014



ECOLE POLYTECHNIQUE

#### Outline

- Motivation: Pedestrian demand management strategies
- 2 Detection: A Bayesian approach for WiFi traces
- 3 Modeling: Path choice in activity network
- 4 Conclusion: Forecasting behavior and building decision-aid tools

(日) (同) (三) (三)

#### Swiss context

By 2030, 100'000 > 25'000 travellers/day between Geneva and Lausanne\* passengers per day \*\*\*\*\* between Geneva and Lausanne > 50'000 travellers/day between Geneva and Lausanne\* \*\*\*\*\*\*\*\*\*\*\*\*\***\*\*\*\*\*** > 100'000 travellers/day between Geneva and Lausanne\* \*

\*\*\*\*\*\*\*\*\*\*\*\*\*\*\*

\* Forecast by Swiss Railways for the maximum scenario

A. Danalet (TRANSP-OR ENAC EPFL)

= 2000 travelers/day

Activity modeling

July 9, 2014 3 / 23

### Pedestrian demand management strategies

- Pedestrian facilities
  - Transportation hubs (train stations, airports, ...)
  - Mass gathering (music festivals, ...)
  - Shops
  - ...
- Challenges
  - Designing efficient buildings
  - Locating points of interest
  - Modifying schedules
  - ...
- $\Rightarrow$  Pedestrian demand management strategies

(人間) トイヨト イヨト

### Activity modeling: Sensitivity to policies



A. Danalet (TRANSP-OR ENAC EPFL)

July 9, 2014 5 / 23

- 一司

#### WiFi traces: No stop, no semantics



A. Danalet (TRANSP-OR ENAC EPFL)

July 9, 2014 6 / 23

#### Generation of activity-episode sequences



A. Danalet (TRANSP-OR ENAC EPFL)

3 July 9, 2014 7 / 23

• = • •

#### Generation of activity-episode sequences



July 9, 2014 8 / 23

.∃ >

### Probabilistic measurement model

$$P(a_{1:K}|\hat{m}_{1:J}) \propto P(\hat{m}_{1:J}|a_{1:K}) \cdot P(a_{1:K})$$

where

- $P(a_{1:K}|\hat{m}_{1:J})$ , the activity probability of an activity-episode sequence
- $P(\hat{m}_{1:J}|a_{1:K}) = \prod_{k=1}^{K} \prod_{j=1}^{J} P(\hat{x}_{j}^{k}|x_{k})$ , the measurement likelihood
- $P(a_{1:K})$ , the prior based on attractivity of the POI

# **Intermediary measurements** Eliminate intermediary measurement if

 $E(t^+) - E(t^-) < T_{min}$ 

since we generate an activity episode at each measurement

#### Sequence elimination procedure



# Individual results



A. Danalet (TRANSP-OR ENAC EPFL)

July 9, 2014 11 / 23

### Aggregate results



July 9, 2014 12 / 23

### Demand analysis

- Model and forecast individual behavior
- Impact on the system



#### (Kirk Anderson)

< ロ > < 同 > < 回 > < 回 > < 回 > < 回

### Observations: activity patterns in a transport hub

Activity types



A. Danalet (TRANSP-OR ENAC EPFL)

### Discrete choice models



- Utility theory : we maximize our satisfaction
- Evaluation of the trade-off between the attributes of the alternatives

- ∢ ∃ ▶

 e.g., willingness to pay (value of time)

# Modeling assumption

- Sequential choice:
  - 1 activity type, sequence, time of day and duration
  - 2 destination choice conditional on 1
- Motivations:
  - Behavior: precedence of activity choice over destination choice
  - Dimensional: destinations  $\times$  time  $\times$  position in the sequence is not tractable

- **(1) ) ) (1) ) ) ) ) )** 

# Activity network

Activity types

Activity network



イロト イ団ト イヨト イヨト

# Activity network



# Challenges

#### Choice set generation

What are the considered alternatives during the choice process?

#### • Utility

What is the mathematical expression of the utility?

#### Correlation structure

Different alternatives share unobserved attributes.

One can get inspired by the route choice literature...

e.g., Metropolis-Hastings algorithm for sampling routes in a network

- 4 同 6 4 日 6 4 日 6

# Utility function

$$V_{\Gamma n} = \eta_k \ln(t_k) + \sum_k \beta_k I_k + \ln \frac{k_{\Gamma n}}{b(\Gamma)}$$

where

•  $\eta_k$  the satiation parameter for activity type k

• 
$$\sum_{k, au} eta_{k, au} I_{k, au}$$
 the time-of-day utility

• 
$$\ln \frac{k_{\Gamma n}}{b(\Gamma)}$$
 is a sampling correction

<ロ> (日) (日) (日) (日) (日)

# Forecasting behavior

• Where are the pedestrians?

WiFi tracking is cheap, covers the whole area and - mixed with other data - is precise enough (Danalet et al.; 2014)

- Why are they here? Parameters of the utility function answer this question (Danalet and Bierlaire; 2014)
- What would happen if some environmental characteristics change? For small variations, the utility function answer this question

(日) (同) (三) (三)

Thank you!

Questions?

A. Danalet (TRANSP-OR ENAC EPFL)

Activity modeling

<mark>১ ১ ৫ ট ১ ট ২ ০ ৫ ৫</mark> July 9, 2014 22 / 23

イロト イヨト イヨト イヨト

#### References I

- Danalet, A. and Bierlaire, M. (2014). A path choice approach to activity modeling with a pedestrian case study, 14th Swiss Transport Research Conference (STRC), Monte Verità, Ascona, Switzerland. URL: http://www.strc.ch/conferences/2014/Danalet\_Bierlaire.pdf
- Danalet, A., Farooq, B. and Bierlaire, M. (2014). A Bayesian approach to detect pedestrian destination-sequences from WiFi signatures, *Transportation Research Part C* 44: 146–170.
  URL: http://dx.doi.org/10.1016/j.trc.2014.03.015
- Lenntorp, B. (1978). A Time-Geographic Simulation Model of Individual Activity Programmes, in T. Carlstein, D. Parkes and N. Thrift (eds), *Timing space and spacing time, vol. 2; Human activity and time* geography, Edward Arn, London, p. 286. URL: http://trid.trb.org/view.aspx?id=1170511

#### Visiosafe in Lausanne train station

#### Animation

January 16, 2013 7h40-7h46

A. Danalet (TRANSP-OR ENAC EPFL)

 July 9, 2014
 ⊇

(日) (同) (三) (三)