

Activity pattern modeling: A path choice approach

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Outline

- 1 Motivation: Pedestrian demand management strategies
- 2 A path choice approach to activity modeling
- 3 Case study: EPFL campus
- 4 Conclusion



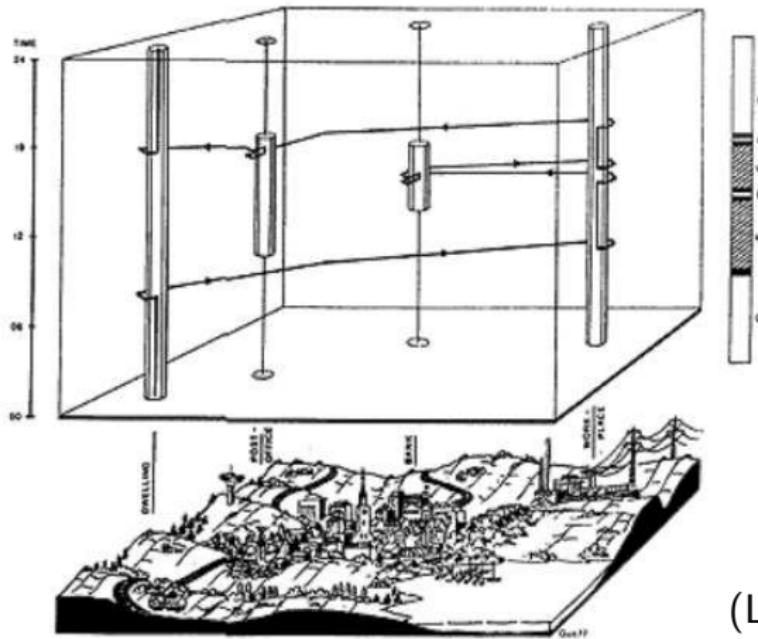
Pedestrian demand management strategies

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

⇒ Pedestrian demand management strategies



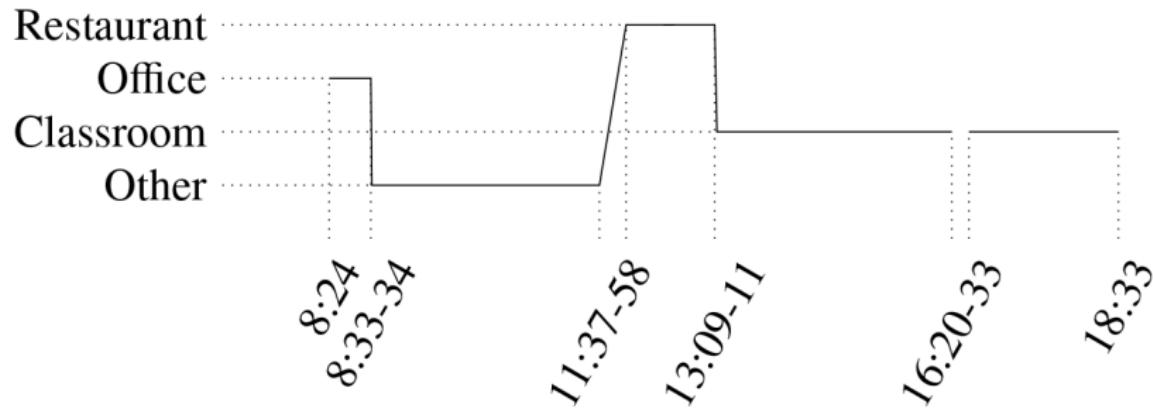
Activity modeling: Sensitivity to policies



(Lenntorp; 1978)

Observations: Activity-episode sequences

Activity types



Modeling assumption

- **Sequential choice:**

- ① activity type, sequence, time of day and duration
- ② destination choice conditional on the previous choice

- Motivations:

- Behavior: “I’m hungry at lunch time”, then “Which restaurant?”
- Dimensional: destinations \times time \times position in the sequence is not tractable

Here we focus on ①.

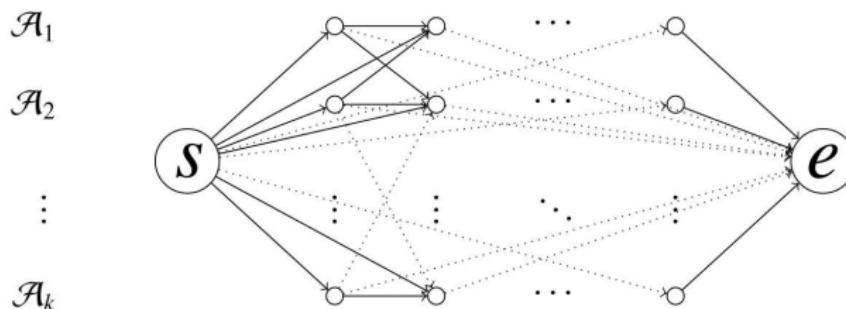
Examples of ②: Ton (2014); Kalakou and Moura (2014).



Activity network

Activity types

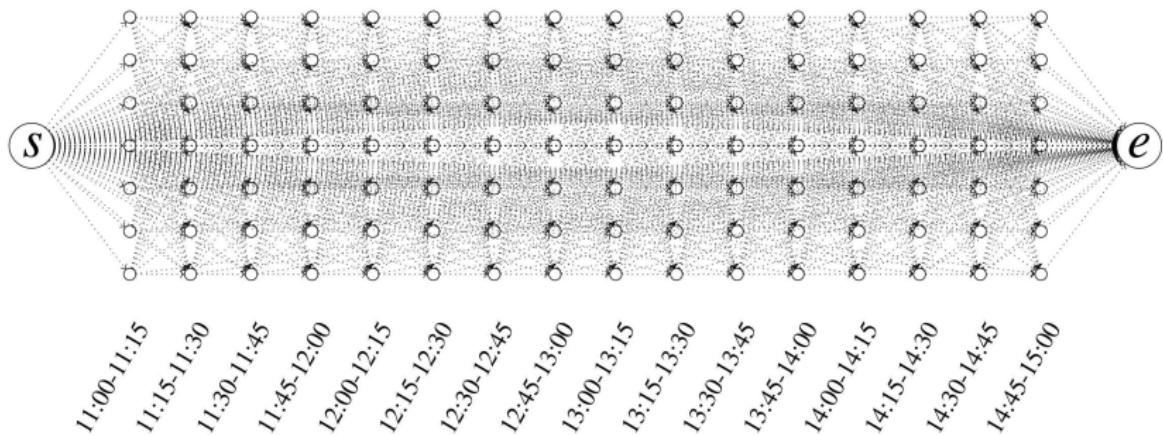
Activity network



1 2 ... T Time

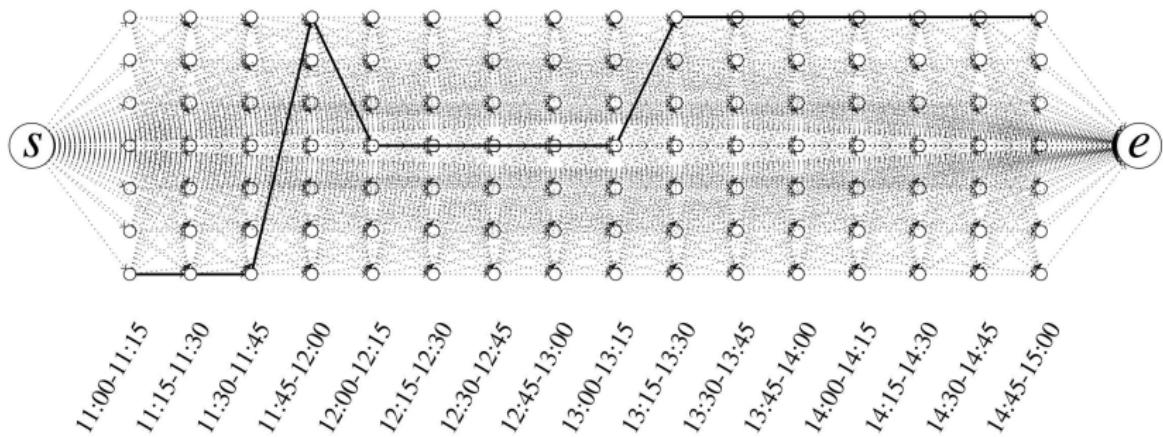
Activity network

Classroom
Shop
Office
Restaurant
Library
Lab
Other



Activity network

Classroom
Shop
Office
Restaurant
Library
Lab
Other

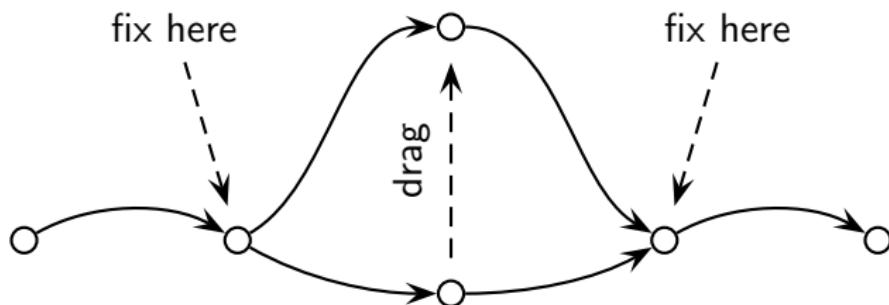


Choice set generation in route choice

- Universal choice set:
 - Too big, not usable
 - Decision maker doesn't consider all routes
- Consideration choice set:
 - Not available
 - Too small
- Sampling of alternatives from the universal choice set:
Metropolis-Hastings algorithm (Flötteröd and Bierlaire; 2013).



Choice set generation: Metropolis-Hastings algorithm



(Flötteröd and Bierlaire; 2013)

Choice set generation in the activity network

- With Metropolis-Hastings algorithm, possibility to define non-link additive cost
- Target weight defined as

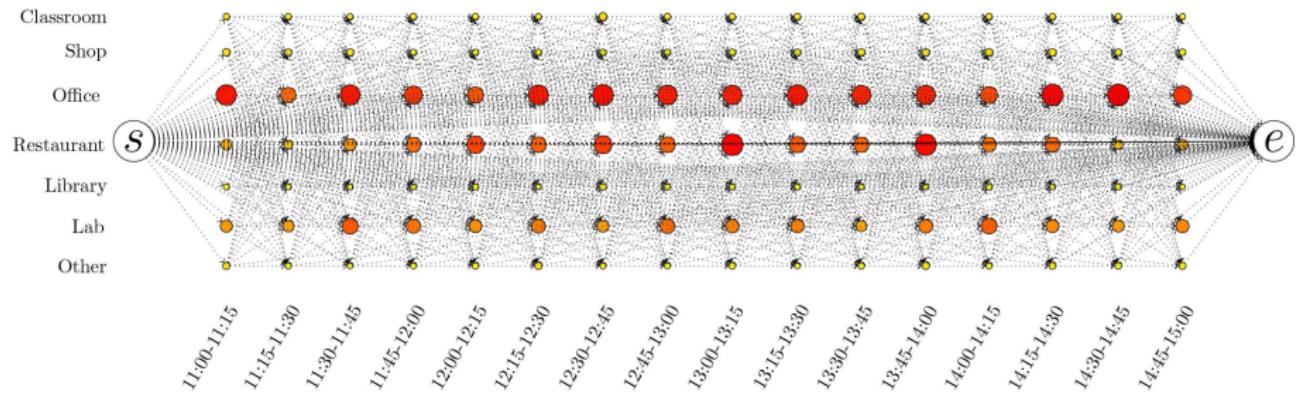
$$\delta(\Gamma) = -\mu_v \cdot \sum_{v \in \Gamma} \delta_v(v) - \delta_\Gamma(\Gamma)$$

with

- link cost: frequency of observations
- path cost: length of observed paths



Activity network: frequency of observations



Activity path choice model for WiFi traces: sampling of alternatives

- Frejinger et al. (2009): a sampling correction must be added

$$\ln q(\mathcal{C}_n | \Gamma) = \ln \frac{k_{\Gamma_n}}{q(\Gamma)}$$

with k_{Γ_n} the number of occurrences and $q(\Gamma)$ the sampling probability

- Sampling probability require full enumeration

$$q(\Gamma) = \frac{b(\Gamma)}{\sum_{\Gamma' \in \mathcal{U}} b(\Gamma')}$$

but cancels out in logit



Activity path choice model for WiFi traces: additive utility function

$$V_T = \sum_{\tau} V(\mathcal{A}_{k,\tau})$$

Inspired by Ettema et al. (2007):

$$V(\mathcal{A}_{k,\tau}) = \eta_k \ln(t_k) + \sum_{k,\tau} \beta_{k,\tau} I_{k,\tau}$$

with

- η_k the satiation parameter for activity type k
- $\sum_{k,\tau} \beta_{k,\tau} I_{k,\tau}$ the time-of-day utility

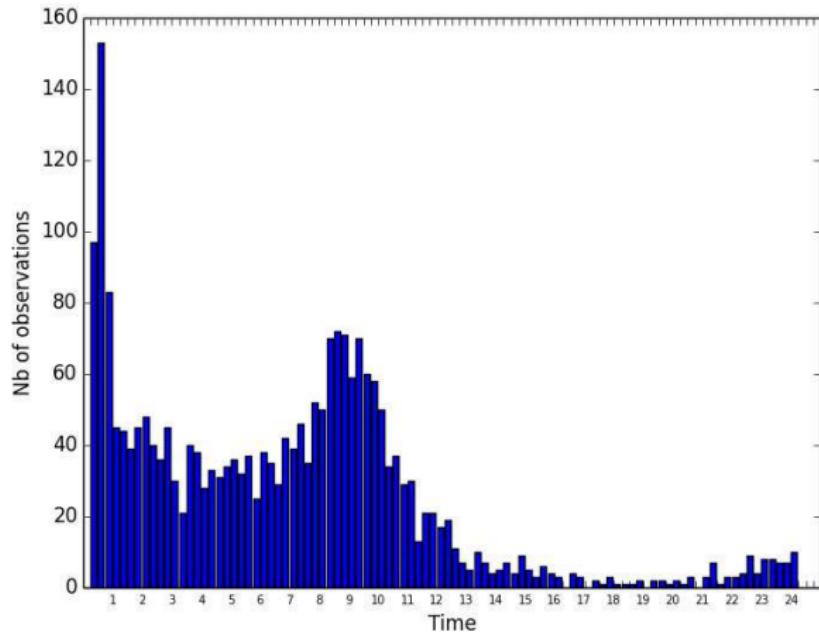


Case study: EPFL campus

- EPFL campus approximately hosts 13'000 people per day
- Similar to transport hubs: some users follow schedules
- Different activities on campus: working, eating, ...
- Recent development of the campus: 1 billion investment in real estate
- WiFi traces processed as in Danalet et al. (2014) with $L = 1$



Length of activity paths

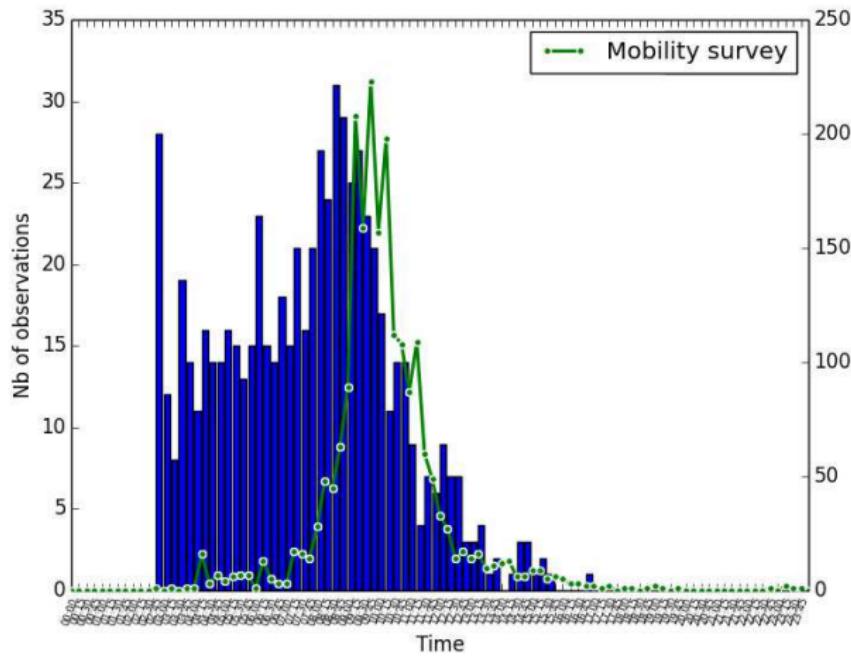


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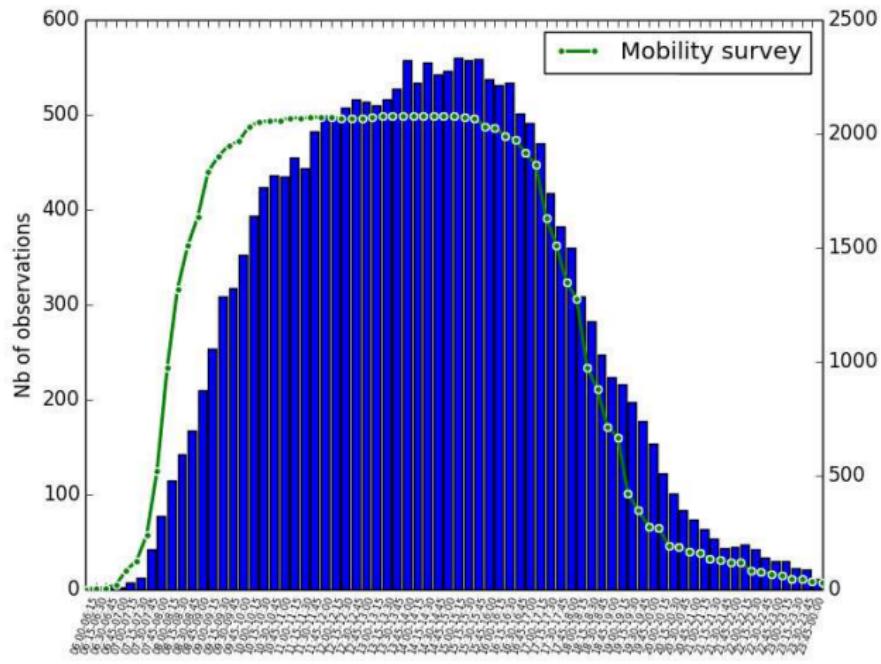
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Length of activity paths: filtering

Removed short activity paths and less than 5.4 measurements / hour.



Time of day



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Estimation of the model

- The model:

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

- Summary statistics:

- Number of alternatives: 1201
- Number of observations = 2219
- Number of estimated parameters = 10

$$\begin{aligned}\mathcal{L}(\beta_0) &= -17952.561 \\ \mathcal{L}(\hat{\beta}) &= -1484.635 \\ -2[\mathcal{L}(\beta_0) - \mathcal{L}(\hat{\beta})] &= 32935.852 \\ \rho^2 &= 0.917 \\ \bar{\rho}^2 &= 0.917\end{aligned}$$

Estimation results for the model

Parameter number	Description	Coeff.	Robust		
		estimate	Asympt. std. error	t-stat	p-value
1	β_{Lab}	-0.337	0.0949	-3.55	0.00
2	β_{Library}	-2.74	0.0795	-34.45	0.00
3	β_{Other}	-2.78	0.0483	-57.62	0.00
4	$\beta_{\text{Restaurant}}$	-0.725	0.0612	-11.85	0.00
5	β_{Shop}	-0.473	0.103	-4.59	0.00
6	η_{Lab}	2.55	0.895	2.85	0.00
7	η_{Office}	-0.787	0.600	-1.31	0.19
8	η_{Other}	9.66	1.27	7.63	0.00
9	$\eta_{\text{Restaurant}}$	5.56	0.789	7.05	0.00
10	$\eta_{\text{Shop, Library, Classroom}}$	-3.26	0.782	-4.16	0.00

Conclusion

- Not home-based, nor tour-based model.
- Can adapt to different contexts and activity types.
- Manage large dimensionality of the problem through importance sampling techniques.
- First results show that the approach is feasible in a realistic context.



Future works

- Improve the specification
 - In particular, add in the utility function variables that are related to the path itself (patterns).
- Evaluate the quality of the generated choice set and its impact on the choice model.
- Correct for the correlation structure
- Manage measurement error



Thank you!

Questions?



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