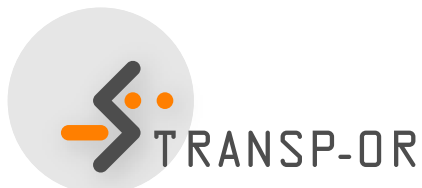


# A latent route choice model in Switzerland

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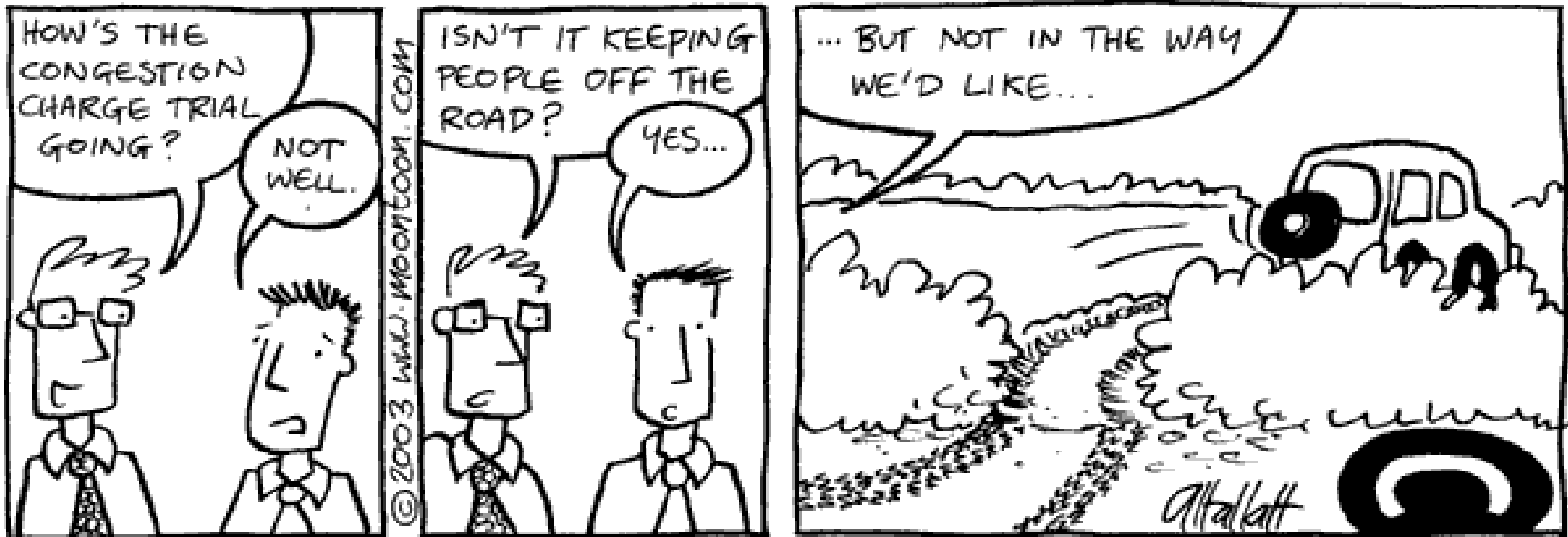


# Outline

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- Swiss mobility pricing project
- Aggregate observations and latent choices
- Modeling approach
- Empirical results
- Conclusion

# Mobility Pricing



# Swiss Mobility Pricing Project

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- A part of a major study on various mobility pricing scenarios in Switzerland
- A collaboration with ETH Zurich and USI Lugano
- Revealed Preferences (RP) and Stated Preferences (SP) data has been collected
- RP data concern long distance route choice by car
  - Route descriptions are approximative
  - Route choices are latent

# Objective

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- Estimate route choice models based on latent chosen routes
- Literature on latent choice models
  - Ben-Akiva et al. (1984), label path approach
  - Ben-Akiva and Lerman (1985), destination choice
  - Toledo et al. (2003), lane choice

# Observations

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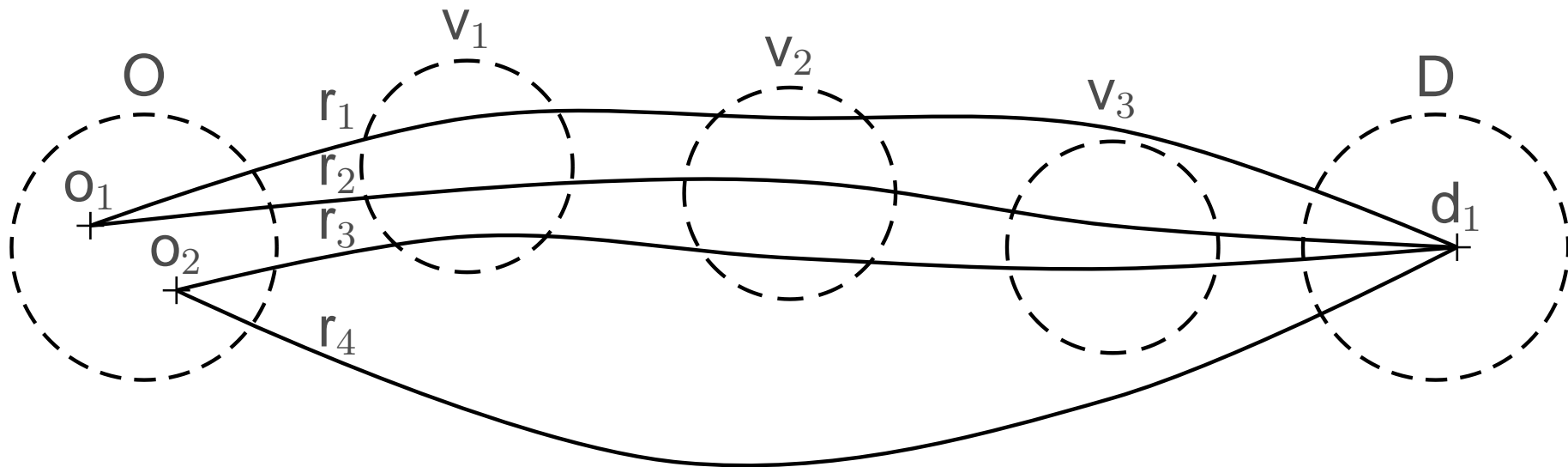
- Exact descriptions of chosen routes are difficult and expensive to obtain
- The concept of path and network as we need for modeling is abstract for respondents
- Here, a chosen route is described by a sequence of cities and locations
- *Aggregate observations* (several paths in the network can correspond to the same observation)

# Observations

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- Better quality of the observations
- Travelers do not need to refer to the network used by the analyst
- Exact origin-destination pairs are not necessarily known
- Exact route is not known

# Observations - Example





# Modeling Approach

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- Several possible modeling approaches
  - Construction of paths from the aggregate observations
    - Involves subjective judgments and generate noise
  - Alternatives in the model are aggregates instead of physical paths
    - Estimated model is of little use in practice
- Our approach: compute the likelihood of an aggregate observation for a classical route choice model

# Modeling Approach

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- Probability of an aggregate observation  $i$ :

$$P(i) = \sum_{s \in S} P(s|i) \sum_{r \in C_s} P(r|i) P(r|C_s)$$

- $s$ : origin-destination pair
- $S$ : set of all origin-destination pairs
- $r$ : route
- $C_s$ : set of all routes for origin-destination pair  $s$

# Modeling Approach

- Probability of an aggregate observation  $i$ :

$$P(i) = \sum_{s \in S} P(s|i) \sum_{r \in C_s} P(r|i) P(r|C_s)$$

- $P(s|i)$  and  $P(r|i)$  can be modeled in several ways
- $P(r|C_s)$ : route choice model that is identifiable if
  1. at least one of the routes in  $C_s$  crosses the observed zones, and
  2. at least one route in  $C_s$  does not cross the observed zones.

- This type of models can be estimated with BIOGEME

# Empirical Results

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- Simplified Swiss network (39411 links and 14841 nodes)
- RP data collection through telephone interviews
- Long distance car travel
- The chosen routes are described with the origin and destination cities as well as 1 to 3 cities or locations that the route pass by
- 940 observations available after data cleaning and verification

# Empirical Results

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# Empirical Results

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- This application is one of few presented in the literature that are based on RP data
- The network is to our knowledge the largest one used for evaluation of route choice modeling approaches

# Empirical Results

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- No information available on the exact origin destination pairs

$$P(s|i) = \frac{1}{|S_i|} \quad \forall s \in S_i$$

- $P(r|i)$  is modeled with a binary variable

$$\delta_{ri} = \begin{cases} 1 & \text{if } r \text{ corresponds to } i \\ 0 & \text{otherwise} \end{cases}$$

# Empirical Results

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- Two origin-destination pairs are randomly chosen for each observation
- 46 routes per choice set are generated with a choice set generation algorithm
- After choice set generation 780 observations are available
  - 160 observations were removed because either all or none of the generated routes crossed the observed zones



# Empirical Results

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- Probability of an aggregate observation  $i$

$$P(i) = \sum_{s \in S_i} \frac{1}{|S_i|} \sum_{r \in C_s} \delta_{ri} P(r|C_s)$$

- We estimate Path Size Logit (Ben-Akiva and Bierlaire, 1999) and Subnetwork (Frejinger and Bierlaire, 2006) models
- BIOGEME ([biogeme.epfl.ch](http://biogeme.epfl.ch)) used for all model estimations

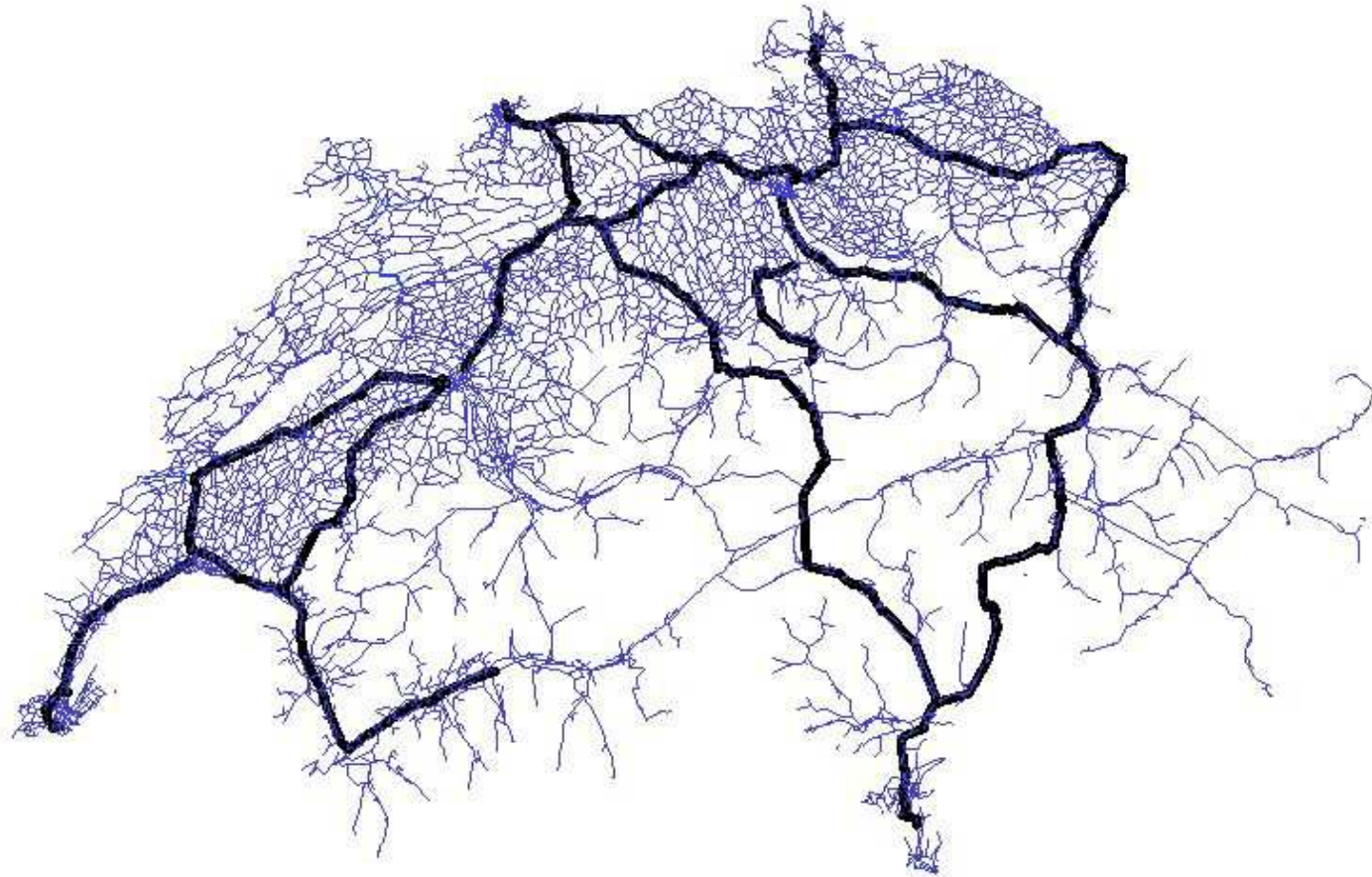
# Empirical Results - Subnetwork

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- Subnetwork: main motorways in Switzerland
- Correlation among routes is explicitly modeled on the subnetwork
- Combined with a Path Size attribute
- Linear-in-parameters utility specifications

# Empirical Results - Subnetwork

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Parameter	PSL		Subnetwork	
<b>In(path size) based on free-flow time</b>	<b>1.04</b>	(0.134) 7.81	<b>1.10</b>	(0.141) 7.78
<i>Scaled Estimate</i>	1.04		1.04	
<b>Freeway free-flow time 0-30 min</b>	<b>-7.12</b>	(0.877) -8.12	<b>-7.45</b>	(0.984) -7.57
<i>Scaled Estimate</i>	-7.12		-7.04	
<b>Freeway free-flow time 30min - 1 hour</b>	<b>-1.69</b>	(0.875) -1.93	<b>-2.26</b>	(1.03) -2.19
<i>Scaled Estimate</i>	-1.69		-2.14	
<b>Freeway free-flow time 1 hour +</b>	<b>-4.98</b>	(0.772) -6.45	<b>-5.64</b>	(1.00) -5.61
<i>Scaled Estimate</i>	-4.98		-5.33	
<b>CN free-flow time 0-30 min</b>	<b>-6.03</b>	(0.882) -6.84	<b>-6.25</b>	(0.975) -6.41
<i>Scaled Estimate</i>	-6.03		-5.91	
<b>CN free-flow time 30 min +</b>	<b>-1.87</b>	(0.331) -5.64	<b>-2.16</b>	(0.384) -5.63
<i>Scaled Estimate</i>	-1.87		-2.04	
<b>Main free-flow travel time 10 min +</b>	<b>-2.03</b>	(0.502) -4.05	<b>-2.46</b>	(0.624) -3.95
<i>Scaled Estimate</i>	-2.03		-2.33	
<b>Small free-flow travel time</b>	<b>-2.16</b>	(0.685) -3.16	<b>-2.75</b>	(0.804) -3.42
<i>Scaled Estimate</i>	-2.16		-2.60	
<b>Proportion of time on freeways</b>	<b>-2.2</b>	(0.812) -2.71	<b>-2.31</b>	(0.865) -2.67
<i>Scaled Estimate</i>	-2.2		-2.18	
<b>Proportion of time on CN</b>	<b>0 fixed</b>		<b>0 fixed</b>	
<b>Proportion of time on main</b>	<b>-4.43</b>	(0.752) -5.88	<b>-4.40</b>	(0.800) -5.51
<i>Scaled Estimate</i>	-4.43		-4.16	
<b>Proportion of time on small</b>	<b>-6.23</b>	(0.992) -6.28	<b>-6.02</b>	(1.03) -5.83
<i>Scaled Estimate</i>	-6.23		-5.69	
<b>Covariance parameter</b>			<b>0.217</b>	(0.0543) 4.00
<i>Scaled Estimate</i>			0.205	

# Empirical Results

	PSL	Subnetwork
Covariance parameter (Rob. Std. Error) Rob. T-test		0.217 (0.0543) 4.00
Number of simulation draws	-	1000
Number of parameters	11	12
Final log-likelihood	-1164.850	-1161.472
Adjusted rho square	0.145	0.147
Sample size: 780, Null log-likelihood: -1375.851		

# Empirical Results

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- All parameters have their expected signs and are significantly different from zero
- The values and significance level are stable across the two models
- The subnetwork model is significantly better than the Path Size Logit (PSL) model

# Conclusion

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- Aggregate observations are convenient to report paths
- They can be used for estimating route choice models
- Care must be taken about the level of aggregation
- Parameters of the RP model are significant and meaningful
- Available in Biogeme / Bioroute