

Demand for public transport services: Integrating qualitative and quantitative methods

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Outline

- 1 Introduction
- 2 Model framework
- 3 Latent variable model
- 4 Integrated model
- 5 Estimation results
- 6 Conclusions

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Introduction

- The research is carried out with CarPostal: bus service in rural and low density areas of Switzerland.
- The aim of the project is to ...
 - Understand the travel behavior in the area of interest
 - Come up with latent attitudes and perceptions affecting travel behavior
 - Integrate these latent attitudes into mode choice context
 - Improve the market share of public transport

Data collection

Data collection campaign consists of the execution of 3 surveys:

- Qualitative survey
- Revealed preferences (RP) survey
- Stated preferences (SP) survey

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 - Identification of potential latent attitudes
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 - Travel diary for a predefined day
 - Socio-economic characteristics
 - Habits
 - Psychometric indicators
- Stated preferences (SP) survey

Data collection

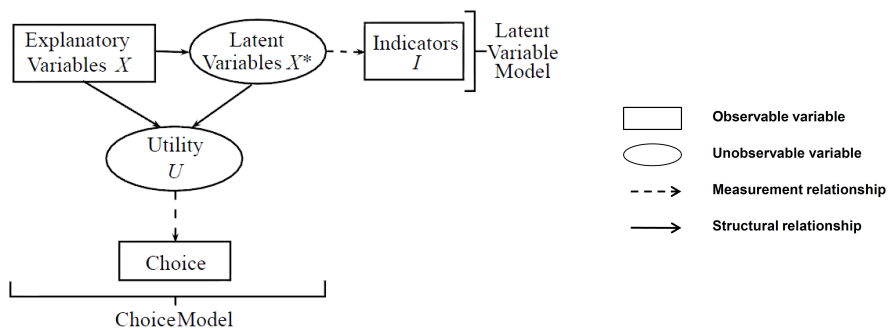
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 - Hypothetical choice situations with improved alternatives:
 - information services, electric bikes, neighborhood service...

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Integrated model framework



Integrated model equations

Structural equations:

- Latent var - explanatory var

$$X_n^* = h(X_n; \lambda) + \omega_n$$

- Utilities

$$U_n = V(X_n, X_n^*; \beta) + \varepsilon_n$$

Measurement equations:

- Latent var - indicators

$$I_n^* = f(X_n^*; \alpha) + \nu$$

- Choice

$$P(y_{in} = 1) = P(U_{in} \geq U_{jn}, \forall j)$$

Likelihood:

$$L_n(y_n, I_n | X_n; \alpha, \beta, \theta_\varepsilon, \theta_\nu, \theta_\omega) =$$

$$\int_{X^*} P(y_n | X_n, X_i, X^*; \beta, \theta_\varepsilon) f(I_n | X_n, X^*; \alpha, \theta_\nu) f(X^* | X_n; \lambda, \theta_\omega) dX^*.$$

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Latent variables

- With the psychometric indicators in RP survey we are able to estimate models with unobserved variables like *attitudes*, *perceptions*, *lifestyle preferences* etc.
- Factor analysis is performed to come up with the most powerful latent factors together with their most explaining indicators. First 6 of them are:
 - Attitude against public transport
 - Environmental concern
 - Public transport awareness
 - Status seeking
 - Pro high density
 - Personalized service

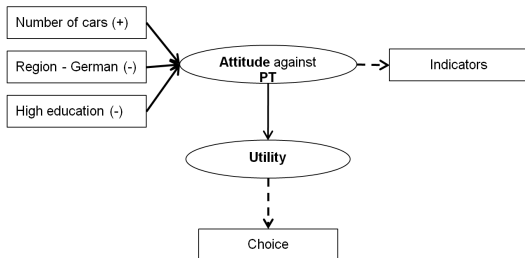
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Attitude against PT

Psychometric indicators (level of agreement: likert scale 1-5)

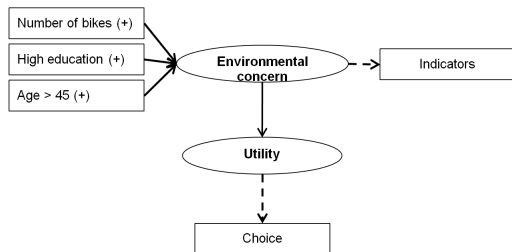
- It is hard to take PT when I travel with my children.
- I do not like to change means of transportation when I travel.
- It is hard to take PT when I have bags or luggage.



Environmental concern

Psychometric indicators (level of agreement: likert scale 1-5)

- I am concerned about global warming.
- We should increase the price of gasoline to reduce congestion and air pollution.
- We must act and make decisions to reduce emissions of greenhouse gases.
- We need more public transport, even if it means higher taxes.



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Integrated latent variable and discrete choice model

- Data from RP survey is used (1124 received questionnaires).
- Concerned choice is the mode choice for each loop of trips in a day (1339 loops).
- There are 3 alternatives for choice:
 - Private mode: car, taxi, car-sharing, moto etc.
 - Public transportation
 - Soft mode: bike and walking

Model specification

$$V_{PM} = ASC_{PM} + \beta_{cost} C_{PM} + \beta_{TT_{PM}} TT_{PM} \\ + \beta_{children} I_{children} + \beta_w I_w + \beta_{french} I_{french}$$

$$V_{PT} = ASC_{PT} + \beta_{cost} C_{PT} + \beta_{TT_{PT}} TT_{PT} \\ + \beta_{freq} F_{PT} + \beta_{attAPt} attAPt + \beta_{attEnv} attEnv$$

$$V_{SM} = ASC_{SM} + \beta_{distance} D_{SM}$$

$$P_i = \frac{\exp(V_i)}{\exp(V_{PM}) + \exp(V_{PT}) + \exp(V_{SM})} \quad i = PM, PT, SM.$$

► Equations for latent variables

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Estimation results - Choice model

		Integrated Model		Multinomial Logit	
Parameter	Utility	Value	t-test	Value	t-test
ASC_{PM}	PM	0.157	0.15*	0.81	3.35
ASC_{PT}	PT	0**	-	0**	-
ASC_{SM}	SM	-0.409	-0.38*	0.218	0.56*
$\beta_{TT_{PM}}$	PM	-0.0211	-4.33	-0.0215	-3.83
$\beta_{TT_{PT}}$	PT	-0.00847	-3.1	-0.00846	-2.79
β_{cost}	PM & PT	-0.0493	-4.63	-0.0508	-3.91
$\beta_{distance}$	SM	-0.221	-4.47	-0.222	-4.44
$\beta_{children}$	PM	0.492	3.09	0.412	2.62
β_w	PM	-0.61	-3.97	-0.622	-4.1
β_{french}	PM	1.05	6.22	1.09	6.5
β_{attAPt}	PT	-0.63	-2.89	-	-
β_{attEnv}	PT	0.326	1.89	-	-
β_{freq}	PT	0.649	3.22	0.701	3.51

(* Statistical significance < 90%, ** Fixed parameter)

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Estimation results - Attitude against PT

Parameter	Value	t-test
β_{attAPt}	-0.63	-2.89
\overline{attAPt}	3.45	54.33
λ_{cars}	0.129	3.52
$\lambda_{high-educ}$	-0.262	5.9
$\lambda_{region-BL}$	-0.307	-3.66
$\lambda_{region-GR}$	-0.234	-2.02
$\lambda_{region-SG}$	-0.315	-3.01
$\lambda_{region-VS}$	-0.193	-2.12
$\lambda_{region-BE}$	-0.467	-3.01

Parameter	Value	t-test
a_{16}	0**	-
a_{17}	0.847	2.28
a_{22}	1.24	4.4
α_{16}	1**	-
α_{17}	0.974	7.84
α_{22}	0.727	7.57
θ_{attAPt}	-0.469	-6.33
θ_{16}	-0.255	-5.03
θ_{17}	-0.126	-3.21
θ_{22}	0.0171	0.71*

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Estimation results - Environmental concern

Parameter	Value	t-test
β_{attEnv}	0.326	1.89
$\frac{\beta_{attEnv}}{attEnv}$	3.04	34.67
$\lambda_{age>45}$	0.00609	2.59
λ_{bikes}	0.0605	4.17
$\lambda_{high-educ}$	0.262	5.9

Parameter	Value	t-test
a_1	-1.77	-2.81
a_2	0.0318	0.07*
a_5	0**	-
a_6	1.06	5
α_1	1.17	6.86
α_2	0.904	7.12
α_5	1**	-
α_6	0.87	15.63
θ_{attEnv}	-0.492	-5.44
θ_1	0.0873	2.6
θ_2	-0.00741	-0.26*
θ_5	-0.174	-3.94
θ_6	-0.582	-12.87

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Results

- Value of time**

	VOT_{PM} (CHF/h)	VOT_{PT} (CHF/h)
Integrated model	25.7	10.3
Multinomial logit	25.4	10.0

- Validation** is done by estimating the model on 80% of the data and predicting the remaining 20%. 66% of the estimated choice probabilities are above 0.5 and 19% are above 0.9.

- Demand elasticities**

	Time elasticity	Cost elasticity
Private mode	-0.20	-0.06
Public transport	-0.34	-0.17

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Conclusions

- With latent variables we are able to have a better understanding of travel behavior.
- When latent attitudes are introduced effect of cost in the utilities is decreased compared to MNL.
- Utility of private mode is explained better compared to MNL since constant becomes insignificant.
- Existence of high education in structural equations of both latent attitudes

Future research

- Inclusion of more than 2 latent variables
- Discrete specification of indicators
- Modeling perceptions as well as attitudes
- Latent classes

- Analysis of SP data
- Model with SP data
- Proposal of improved alternatives

Thank you for your attention !



Equations for latent variables

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Structural equations for latent variables:

$$Att = \overline{Att} + \sum_e \lambda_e X_e + \omega,$$

Measurement equations for latent variables:

$$I_k = a_k + \alpha_k Att + v_k \quad \forall k,$$