
ENHANCED MEASUREMENT EQUATIONS FOR LATENT CLASS CHOICE MODELS

Aurélie Glerum
Ricardo Hurtubia
My Hang Nguyen
Michel Bierlaire

Eighth Workshop on Discrete Choice Models

Friday 31st May 2013

Introduction & motivation

Literature

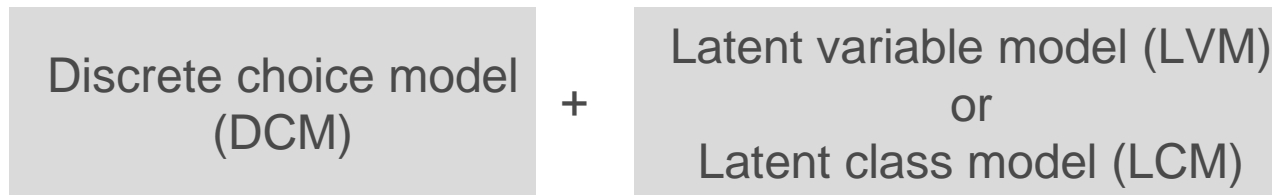
Methodology

Mode choice case study

Conclusion

Recent developments in demand modeling for transportation

- **Hybrid choice model (HCM)** framework (Walker, 2001; Ben-Akiva et al., 2002)
Comprehensive framework that allows to incorporate unobservable factors as explanatory variables of choice.



- Choice of transportation mode, car, etc.
 - Influenced by economic factors:
 - Price
 - Trip duration
 - Etc.
 - Often also involve more subjective factors:
 - Attitudes
 - Perceptions
 - Lifestyles
 - Habits
- HCM framework incorporates these subjective factors.

Important issues in the use of HCMs:

- Well-established that soft features have an important effect on choice.
- The best way to integrate them in choice models is not obvious.
- Many applications using psychometrics as measures of attitudes in the integrated choice and latent variable model framework, fewer for latent class choice models.

Aim of this research:

- Enhance latent class choice models with psychometric indicators

Integration of psychometrics in latent class choice models:

- Measurement indicators usually used for LVM, but relatively few applications using them in LCM.
 - Specification of the class-membership model of a LC by means of an intermediate LV (Gopinath, 1995).
 - Estimation of class-specific probabilities to respond to each indicator (Collins and Lanza, 2010; Atasoy et al., forthcoming).

Integration of psychometrics in latent class choice models:

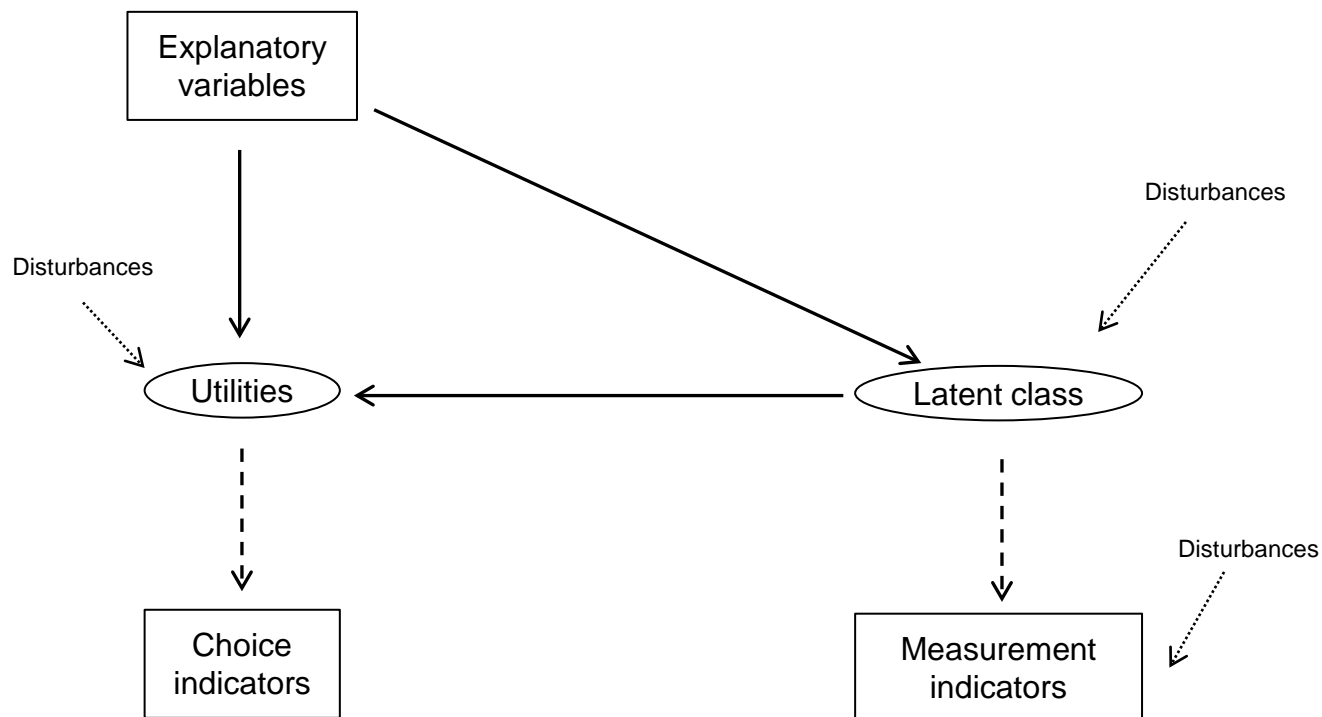
- Measurement indicators usually used for LVM, but relatively few applications using them in LCM.
- Specification of the class-membership model of a LC by means of an intermediate LV (Gopinath, 1995).

AIM \Rightarrow **Simpler characterisation where direct link between indicators and latent classes.**

- Estimation of class-specific probabilities to respond to each indicator (Collins and Lanza, 2010; Atasoy et al., forthcoming).

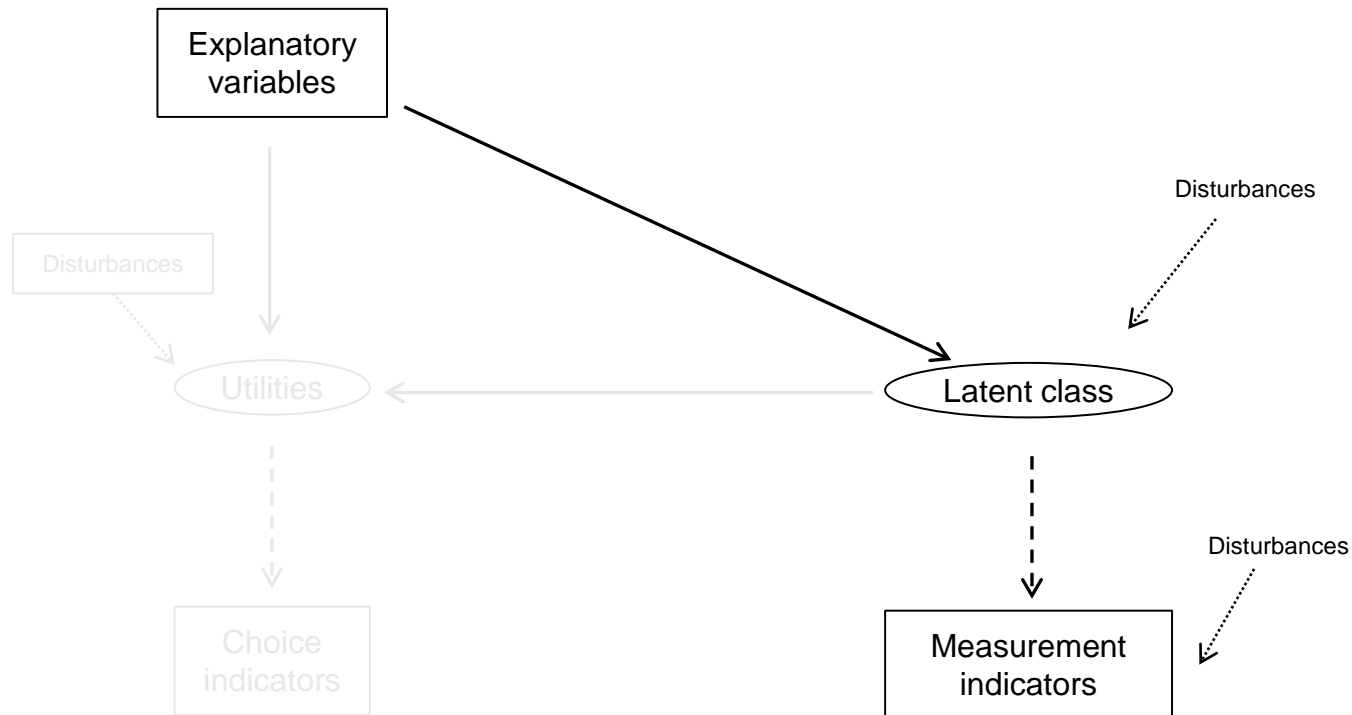
AIM \Rightarrow **Add structural information in the measurement equation**

Integration of psychometrics in latent class choice models:



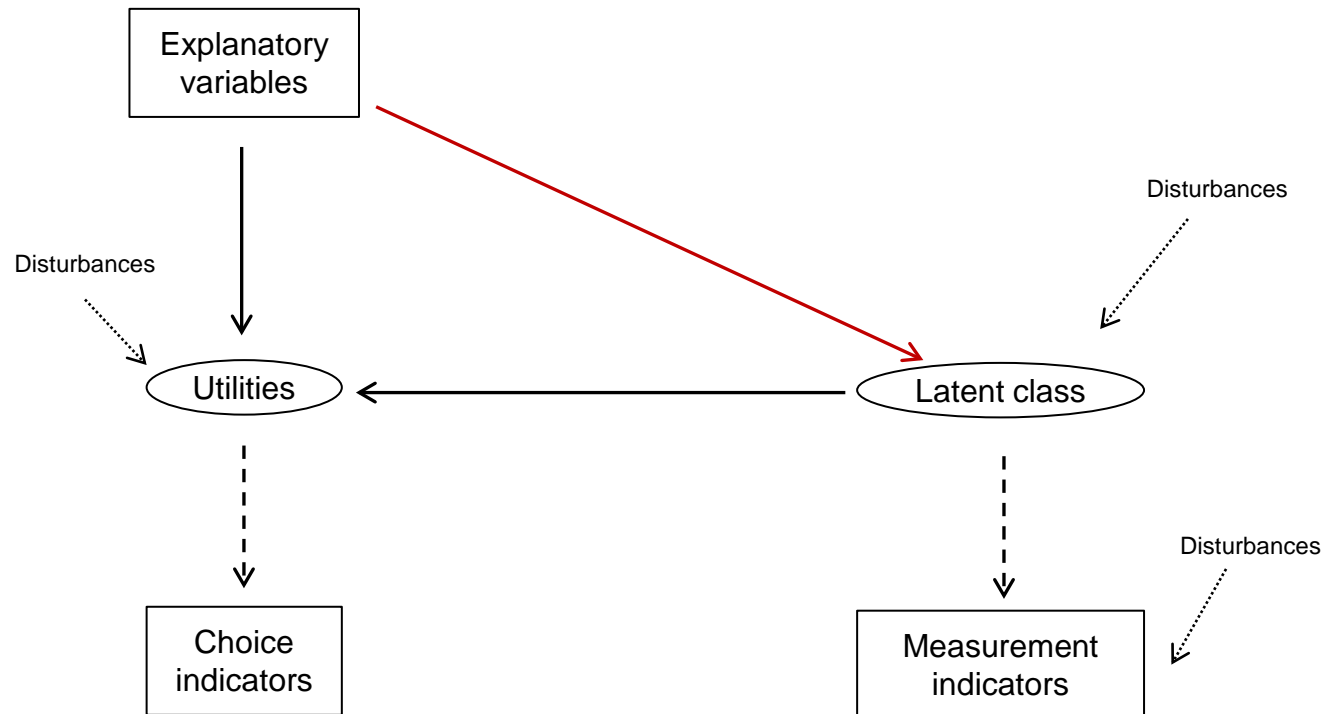
Integration of psychometrics in latent class choice models:

- **Structural equation model (SEM)** framework used to characterize latent construct and relate it to its measurement indicators
(e.g. Bollen, 1989; Hancock and Mueller, 2006; Bartholomew et al., 2011).



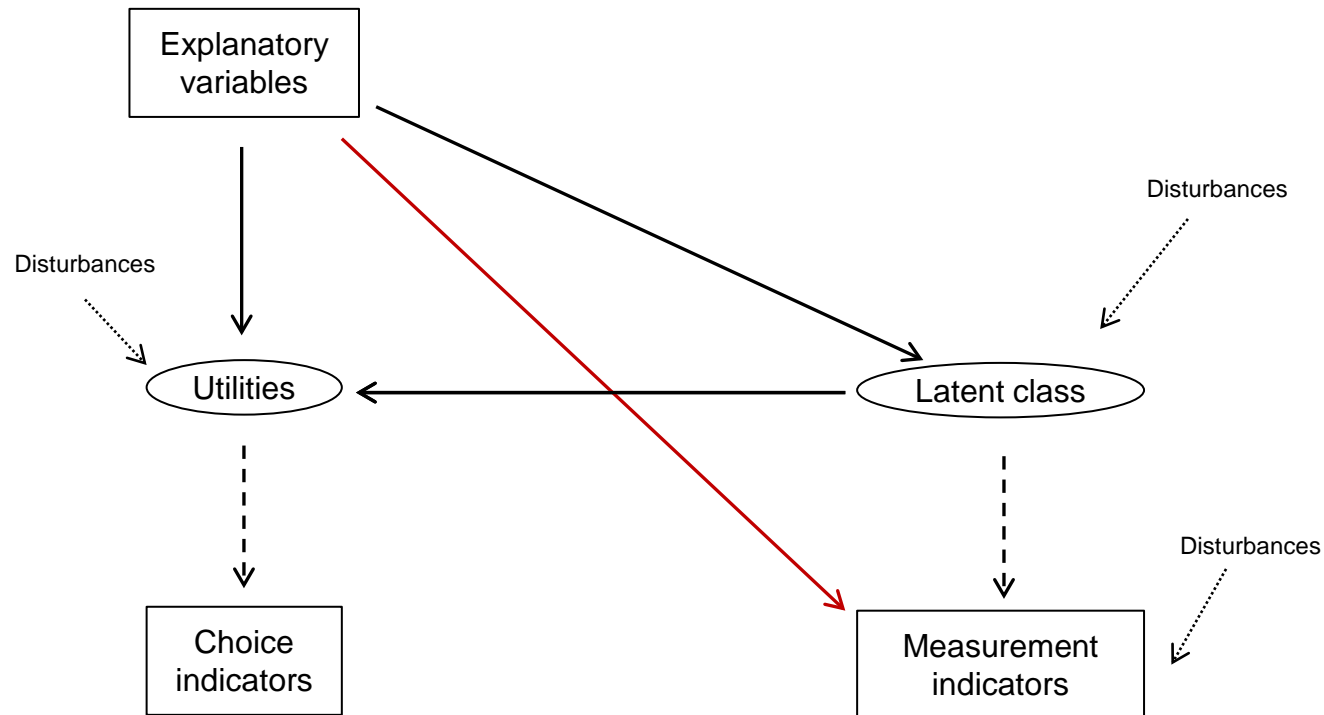
Integration of psychometrics in latent class choice models:

- Heterogeneity of membership to latent classes captured among population
- But also need to capture heterogeneity in reporting indicators of latent class



Integration of psychometrics in latent class choice models:

- Heterogeneity of membership to latent classes captured among population
- **But also need to capture heterogeneity in reporting indicators of latent class**



Specification of integrated choice and latent class model

Choice model:

$$U_{in}^s = V(X_{in}, X_n, \beta^s) + \varepsilon_{in}^s, \quad \text{with } \varepsilon_{in}^s \sim EV(0, 1)$$

Latent class model:

Class-membership function:

$$F_{ns} = f(X_n, \gamma^s) + \xi_{ns}, \quad \text{where } \xi_{ns} \sim EV(0, 1)$$

Measurement indicators:

$$\tilde{G}_{I_k, n}^s = g(X_n; \alpha_k^s) + v_{kn}^s,$$

$$\text{with } v_{kn}^s \sim \text{Logistic}(0, 1)$$

$$G_{I_k, n}^s = \begin{cases} 1 & \text{if } -\infty < \tilde{G}_{I_k, n}^s \leq \tau_{1, k}^s \\ 2 & \text{if } \tau_{1, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{2, k}^s \\ 3 & \text{if } \tau_{2, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{3, k}^s \\ 4 & \text{if } \tau_{3, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{4, k}^s \\ 5 & \text{if } \tau_{4, k}^s < \tilde{G}_{I_k, n}^s \leq +\infty \end{cases}$$

Specification of integrated choice and latent class model

Choice model:

$$U_{in}^s = V(X_{in}, X_n, \beta^s) + \varepsilon_{in}^s, \quad \text{with } \varepsilon_{in}^s \sim EV(0, 1)$$

Latent class model:

Class-membership function:

$$F_{ns} = f(X_n, \gamma^s) + \xi_{ns}, \quad \text{where } \xi_{ns} \sim EV(0, 1)$$

Measurement indicators:

$$\tilde{G}_{I_k, n}^s = g(X_n, \alpha_k^s) + v_{kn}^s,$$

with $v_{kn}^s \sim \text{Logistic}(0, 1)$

↑
Class-specific
parameters

$$G_{I_k, n}^s = \begin{cases} 1 & \text{if } -\infty < \tilde{G}_{I_k, n}^s \leq \tau_{1, k}^s \\ 2 & \text{if } \tau_{1, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{2, k}^s \\ 3 & \text{if } \tau_{2, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{3, k}^s \\ 4 & \text{if } \tau_{3, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{4, k}^s \\ 5 & \text{if } \tau_{4, k}^s < \tilde{G}_{I_k, n}^s \leq +\infty \end{cases}$$

Specification of integrated choice and latent class model

Choice model:

$$U_{in}^s = V(X_{in}, X_n, \beta^s) + \varepsilon_{in}^s, \quad \text{with } \varepsilon_{in}^s \sim EV(0, 1)$$

Latent class model:

Class-membership function:

$$F_{ns} = f(X_n, \gamma^s) + \xi_{ns}, \quad \text{where } \xi_{ns} \sim EV(0, 1)$$

Measurement indicators:

$$\tilde{G}_{I_k, n}^s = g(X_n, \alpha_k^s) + v_{kn}^s,$$

with $v_{kn}^s \sim \text{Logistic}(0, 1)$

Socio-economic information
of the individual

$$G_{I_k, n}^s = \begin{cases} 1 & \text{if } -\infty < \tilde{G}_{I_k, n}^s \leq \tau_{1, k}^s \\ 2 & \text{if } \tau_{1, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{2, k}^s \\ 3 & \text{if } \tau_{2, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{3, k}^s \\ 4 & \text{if } \tau_{3, k}^s < \tilde{G}_{I_k, n}^s \leq \tau_{4, k}^s \\ 5 & \text{if } \tau_{4, k}^s < \tilde{G}_{I_k, n}^s \leq +\infty \end{cases}$$

Specification of integrated choice and latent class model

Likelihood function given by: $L = \prod_{n=1}^N f(y_{in}, I_n | X_{in}; \alpha, \beta, \lambda)$ with

$$f(y_{in}, I_n | X_{in}, s; \alpha, \beta, \lambda) = \sum_s \left\{ P_n(y_{in} | X_{in}, s; \beta) \prod_k P(I_{kn} | X_{in}, s; \alpha_s) \right\} P_n(s | X_n, s; \lambda)$$

$$y_{in} = \begin{cases} 1 & \text{if } U_{in} = \max_j U_{jn} \\ 0 & \text{otherwise} \end{cases}$$

Specification of integrated choice and latent class model

Likelihood function given by: $L = \prod_{n=1}^N f(y_{in}, I_n | X_{in}; \alpha, \beta, \lambda)$ with

$$f(y_{in}, I_n | X_{in}, s; \alpha, \beta, \lambda) = \sum_s \left\{ P_n(y_{in} | X_{in}, s; \beta) \prod_k P(I_{kn} | X_{in}, s; \alpha_s) \right\} P_n(s | X_n, s; \lambda)$$

$$y_{in} = \begin{cases} 1 & \text{if } U_{in} = \max_j U_{jn} \\ 0 & \text{otherwise} \end{cases}$$

⇒ **Analogy with integrated choice and latent variable model**

$$f(y_{in}, I_n | X_{in}; \alpha, \beta, \lambda) = \int_{X_n^*} P(y_{in} | X_{in}, X_n^*; \beta) \cdot \prod_k P(I_{kn} | X_{in}, X_n^*; \alpha) \cdot f(X_n^* | X_n; \lambda) dX_n^*$$

Specification of integrated choice and latent class model

Likelihood function given by: $L = \prod_{n=1}^N f(y_{in}, I_n | X_{in}; \alpha, \beta, \lambda)$ with

$$f(y_{in}, I_n | X_{in}, s; \alpha, \beta, \lambda) = \sum_s \left\{ P_n(y_{in} | X_{in}, s; \beta) \prod_k P(I_{kn} | X_{in}, s; \alpha_s) \right\} P_n(s | X_n, s; \lambda)$$

$$y_{in} = \begin{cases} 1 & \text{if } U_{in} = \max_j U_{jn} \\ 0 & \text{otherwise} \end{cases}$$

⇒ Analogy with integrated choice and latent variable model

⇒ Class-specific parameters in measurement equation

Specification of integrated choice and latent class model

Likelihood function given by: $L = \prod_{n=1}^N f(y_{in}, I_n | X_{in}; \alpha, \beta, \lambda)$ with

$$f(y_{in}, I_n | X_{in}, s; \alpha, \beta, \lambda) = \sum_s \left\{ P_n(y_{in} | X_{in}, s; \beta) \prod_k P(I_{kn} | X_{in}, s; \alpha_s) \right\} P_n(s | X_n, s; \lambda)$$

$$y_{in} = \begin{cases} 1 & \text{if } U_{in} = \max_j U_{jn} \\ 0 & \text{otherwise} \end{cases}$$

- ⇒ Analogy with integrated choice and latent variable model
- ⇒ Class-specific parameters in measurement equation
- ⇒ Socio-economic variables as explanatory of response to indicator

Specification of integrated choice and latent class model

Likelihood function given by: $L = \prod_{n=1}^N f(y_{in}, I_n | X_{in}; \alpha, \beta, \lambda)$ with

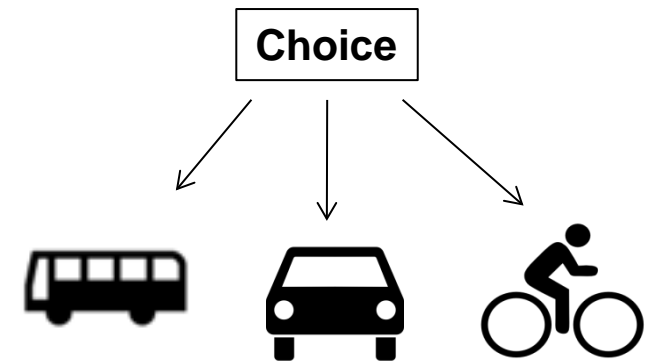
$$f(y_{in}, I_n | X_{in}, s; \alpha, \beta, \lambda) = \sum_s \left\{ P_n(y_{in} | X_{in}, s; \beta) \prod_k P(I_{kn} | X_{in}, s; \alpha_s) \right\} P_n(s | X_n, s; \lambda)$$

$$y_{in} = \begin{cases} 1 & \text{if } U_{in} = \max_j U_{jn} \\ 0 & \text{otherwise} \end{cases}$$



- ⇒ Analogy with integrated choice and latent variable model
- ⇒ Class-specific parameters in measurement equation
- ⇒ Socio-economic variables as explanatory of response to indicator

Revealed preferences (RP) survey

- **Transportation mode choice study**
- Conducted between 2009-2010 in low-density areas of Switzerland
- Conducted with bus operator PostBus
- Info on **all round trips performed by inhabitants in one day**:
 - Transport mode
 - Trip duration
 - Cost of trip
 - Activity at destination
 - Etc.
- Psychometric indicators
- 1763 valid questionnaires collected



Development of latent class choice model:

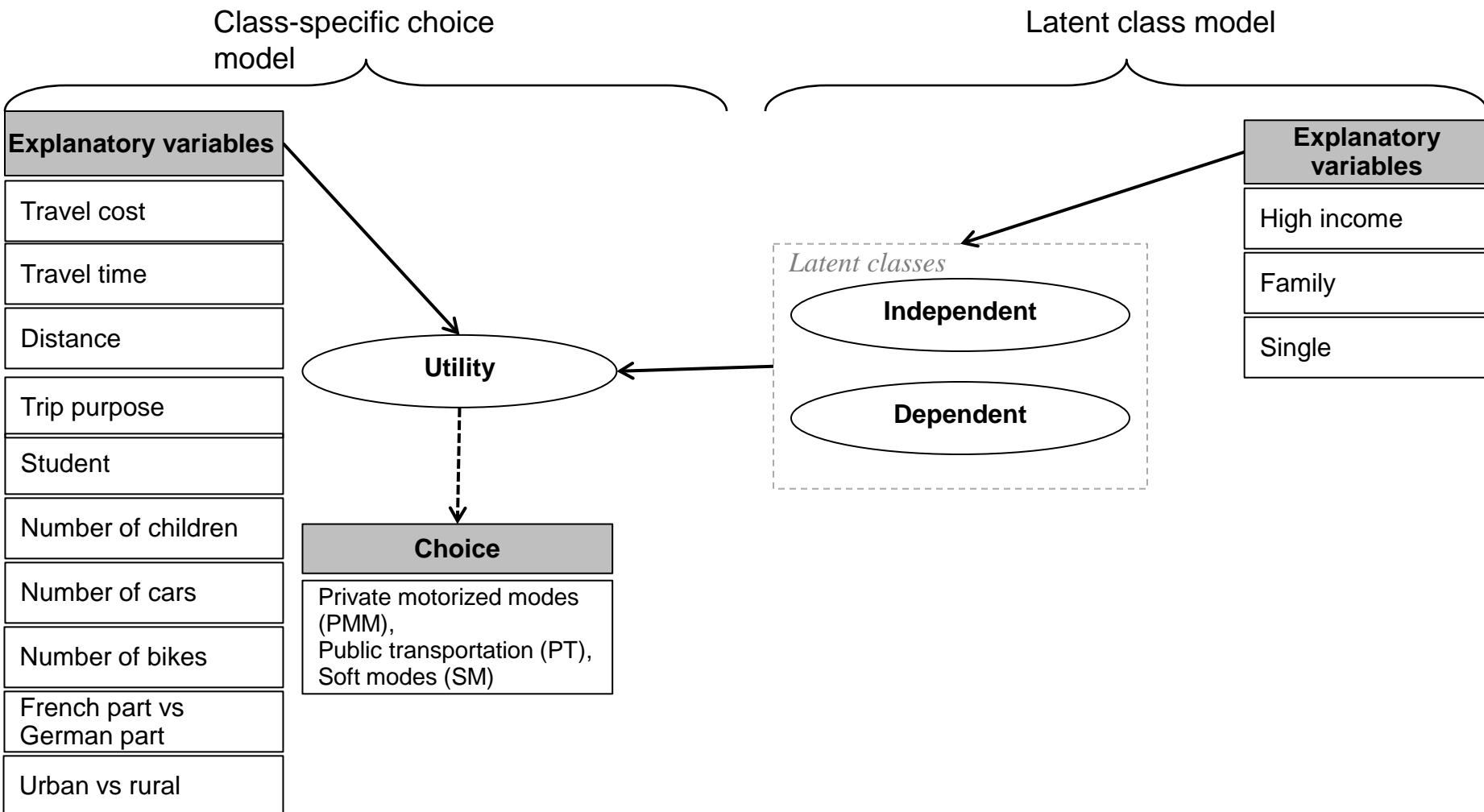
Analysis of the transportation mode choices for individuals segmented according to  dependent class
 independent class

Comparison of 3 models:

LCCM 1: without psychometric indicators

LCCM 2: with psychometric indicators, estimation of class-specific item-response probabilities (Atasoy et al., 2012)

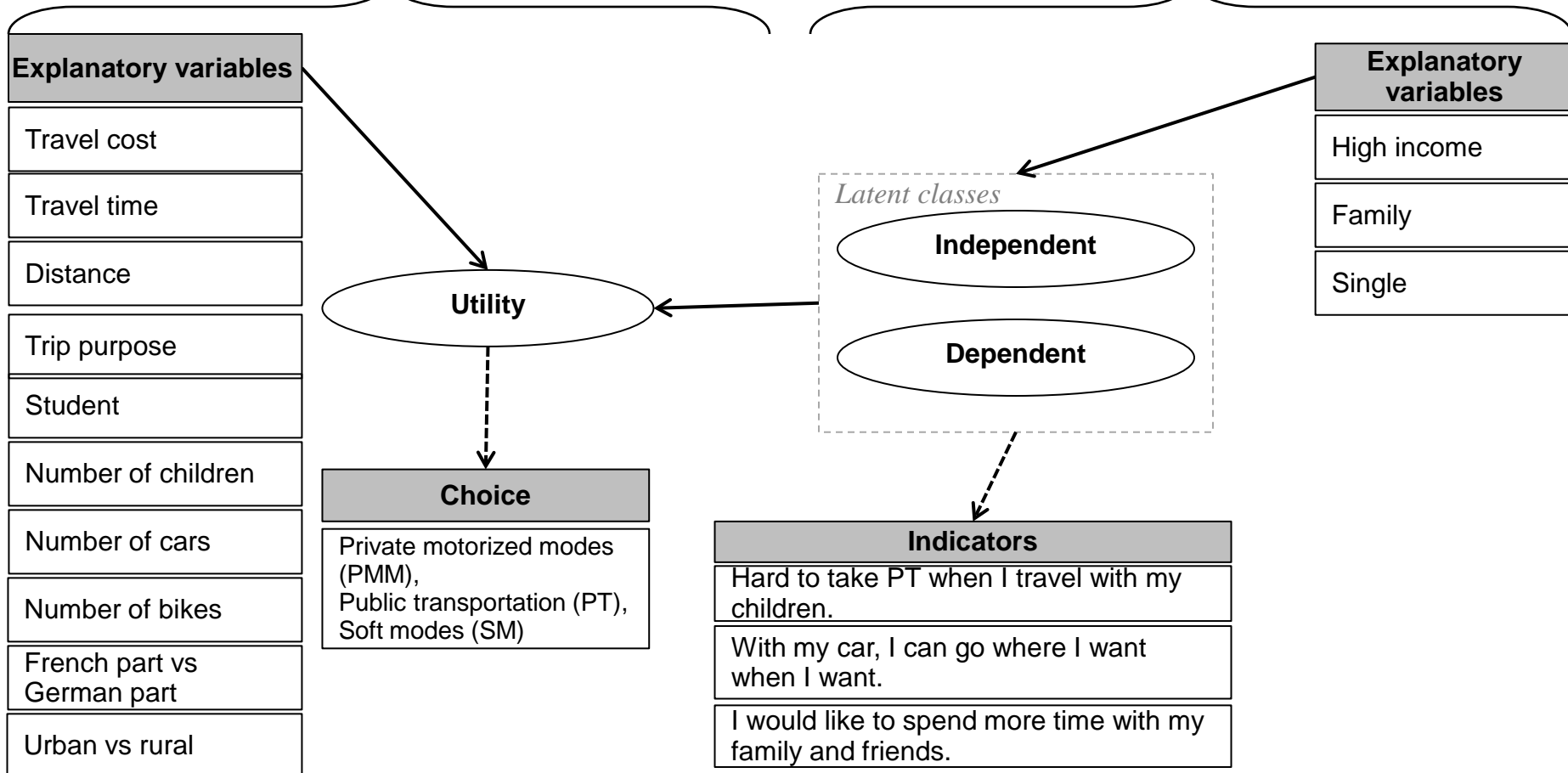
LCCM 3: with psychometric indicators, related to socio-economic characteristics of the respondent by structural measurement relationships



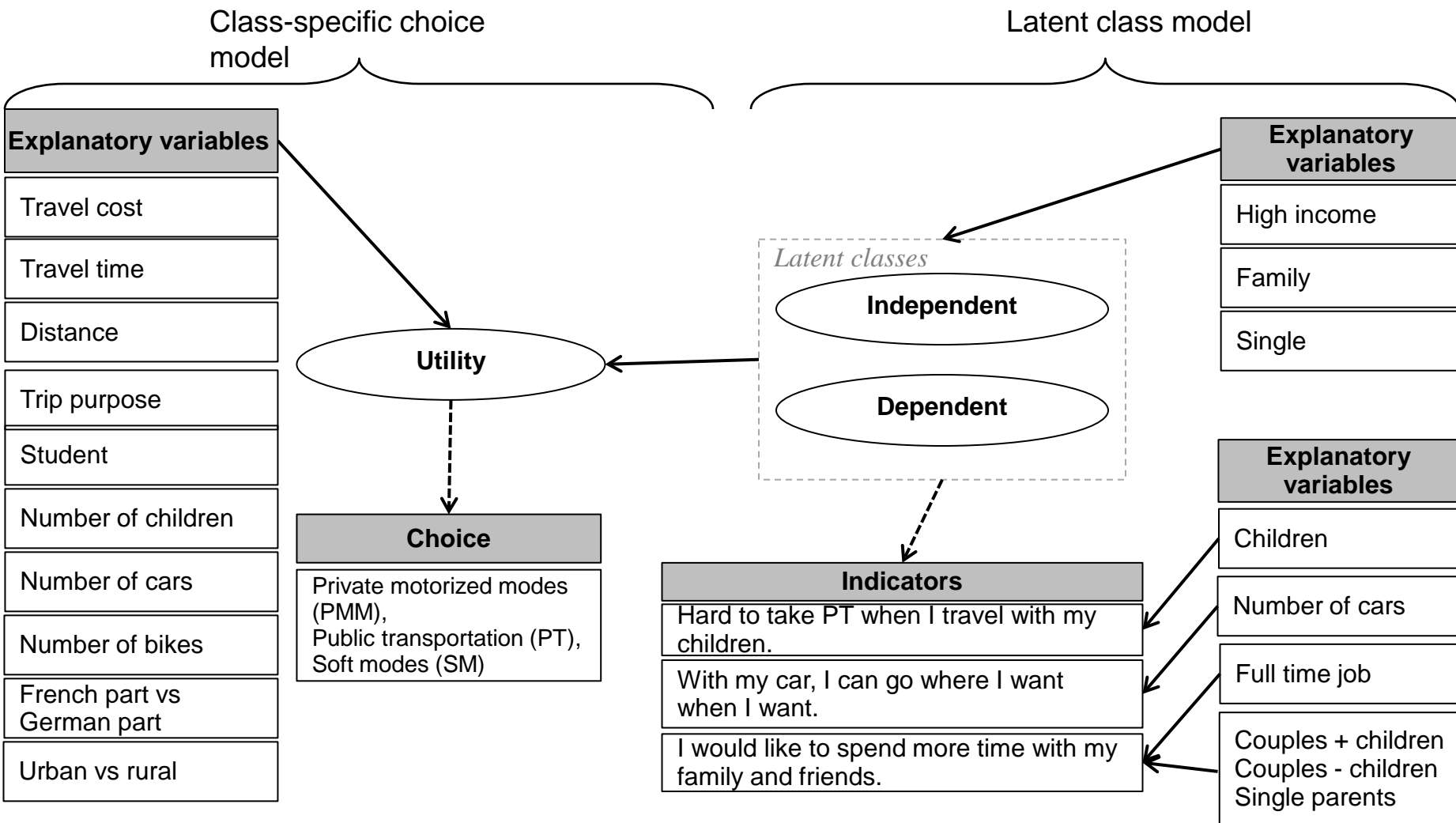
$$\mathcal{L}_{LCM1} = \prod_n \{P_n(i|\text{class 1}) \cdot P_n(\text{class 1}) + P_n(i|\text{class 2}) \cdot P_n(\text{class 2})\}$$

Class-specific choice model

Latent class model



$$\mathcal{L}_{LCM2} = \prod_n \{ P_n(i|\text{class 1}) \cdot \pi_{11} \cdot \pi_{21} \cdot \pi_{31} \cdot P_n(\text{class 1}) + P_n(i|\text{class 2}) \cdot \pi_{12} \cdot \pi_{22} \cdot \pi_{32} \cdot P_n(\text{class 2}) \}$$



$$\mathcal{L}_{LCM3} = \prod_n \{ P_n(i|\text{class } 1) \cdot P_n(I1|\text{class } 1) \cdot P_n(I2|\text{class } 1) \cdot P_n(I3|\text{class } 1) \cdot P_n(\text{class } 1) + P_n(i|\text{class } 2) \cdot P_n(I1|\text{class } 2) \cdot P_n(I2|\text{class } 2) \cdot P_n(I3|\text{class } 2) \cdot P_n(\text{class } 2) \}$$

Estimation results for the class-membership model

	LCCM 1		LCCM 2		LCCM 3	
Parameters	estimate	t-test	estimate	t-test	estimate	t-test
ASC_{class}	-0.215	-0.86**	-0.629	-3.25	-0.589	-3.39
γ_{family}	0.136	0.51**	3.92	4.84	0.967	5.41
γ_{income}	0.693	2.76	0.460	2.22	0.684	4.50
γ_{single}	0.408	1.34**	0.704	3.57	0.743	3.33

- **Increase of the significance** of the parameters of the latent class model.

Estimation results for the class-membership model

	LCCM 1		LCCM 2		LCCM 3	
Parameters	estimate	t-test	estimate	t-test	estimate	t-test
ASC_{class}	-0.215	-0.86**	-0.629	-3.25	-0.589	-3.39
γ_{family}	0.136	0.51**	3.92	4.84	0.967	5.41
γ_{income}	0.693	2.76	0.460	2.22	0.684	4.50
γ_{single}	0.408	1.34**	0.704	3.57	0.743	3.33

- Increase of the significance of the parameters of the latent class model.
- **Income** parameter has become **more important**.

Model application: computation of VOT

		VOT PMM [CHF/hour]	VOT PT [CHF/hour]
LCCM 1	Class independent	3.06	3.72
	Class dependent	52.63	17.53
	Overall	28.97	10.94
LCCM 2	Class independent	35.78	15.38
	Class dependent	22.05	8.84
	Overall	29.53	12.40
LCCM 3	Class independent	63.27	16.21
	Class dependent	34.16	5.99
	Overall	36.94	18.40

- VOTs comparable with literature on transport economics (Jara-Diaz, 2007), where VOT can be compared to wage rate.
- Individuals in the independent class have higher incomes (> 8000 CHF), hence a higher value of time.

Main findings:

- Specified **LCCM with psychometric indicators as measurements of LC**
- Account for **heterogeneity of response behavior** that can be captured by individual-specific information in measurement model of LCM
- **Importance of accounting for it in LCCM**
 - Socio-economic characteristics affect response to opinion questions significantly
 - Parameters of the class membership utility increase in significance
 - VOT are comparable with existing studies
- **Residual analysis needed to further validate the model**

Thanks!