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# Behavioral Models for Impact Assessment of Autonomous Driving on Travel Demand

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## INTRODUCTION

### **Background Concepts**

"Autonomous Vehicles (AVs) are means of transport that are capable of *sensing* the environment and *moving* safely without human intervention"

#### **Potential impacts on urban sub-systems** Infrastructure capacity (e.g. highways, parking spaces) **Transportation** Mobility choices (e.g. ownership or consumption?) Travel behaviors (e.g. individual or collective use?) Accessibility Land Use Residential and Business location choices Urban form (e.g. urban sprawl) Local emissions (i.e. pollutants) Environment Global emissions (i.e. greenhouse gases) Energy consumption IN-CAR DISPLAT Employment ٠ Economy Real estate price Quality of life WAYMO 01 Society Equity Road safety and cybersecurity

#### **References:**

- Milakis D., van Arem B., van Wee B., 2017. Policy and society related implications of automated driving. Journal of Intelligent Transportation Systems, 2017, Vol. 21, No. 4, 324-348.
- Duarte F., Ratti C., 2018. The Impact of Autonomous Vehicles on Cities: A Review. J. Urban Technol. 25, 3–18.
- Coppola P., Silvestri F., 2019. Autonomous vehicles and future mobility solutions. In Autonomous Vehicles and Future Mobility, AET series, Elsevier, ISBN: 978-0-12-817696-2.
- Silvestri F., De Fabiis F., Coppola P., 2022. Veicoli a guida autonoma e mobilità post-car. In Urban@It, Ottavo Rapporto sulle città, ISBN: 978-88-15-38276-4.



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## INTRODUCTION

### **Demand-related Uncertainty Factors**



#### **References:**

- Davis F.D., 1989. Perceived Usefulness, Perceived Ease of Use, and User Acceptance of Information Technology. MIS Q. 13, 319–340.
- Ajzen I., 1991. The theory of planned behavior. Organ. Behav. Hum. Decis. Process., Theories of Cognitive Self-Regulation 50, 179–211.
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- Nastjuk I., Herrenkind B., Marrone M., Brendel A.B., Kolbe L.M., 2020. What drives the acceptance of autonomous driving? Technol. Forecast. Soc. Change 161, 120319.



# LITERATURE REVIEW

## **Autonomous Driving & Travel Demand Modeling**



- Krueger R., Rashidi T.H., Rose J.M., 2016. Preferences for shared autonomous vehicles. Transportation Research Part C, 69, 343–355.
- Winter K., Cats O., Martens K., van Arem B., 2020. Identifying user classes for shared and automated mobility services. European Transport Research Review 12, 36.
- Asgari H., Jin X., Corkery T., 2018. A Stated Preference Survey Approach to Understanding Mobility Choices in Light of Shared Mobility Services and Automated Vehicle Technologies in the U.S. Transportation Research Record 2672, 12–22.
- Stoiber T., Schubert I., Hoerler R., Burger P., 2019. Will consumers prefer shared and pooled-use autonomous vehicles? Transportation Research Part D, 71, 265–282.
- Lavieri P.S., Bhat C.R., 2019. Modeling individuals' willingness to share trips with strangers in an autonomous vehicle future. Transportation Research Part A, 124, 242–261.





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#### **References:**

- Haboucha C.J., Ishaq R., Shiftan Y., 2017. User preferences regarding autonomous vehicles. Transportation Research Part C, 78, 37–49.
- Chee P.N.E., Susilo Y.O., Wong Y.D., 2020. Determinants of intention-to-use first-/last-mile automated bus service. Transportation Research Part A, 139, 350–375.
- Acheampong R.A., Cugurullo F., 2019. Capturing the behavioural determinants behind the adoption of autonomous vehicles. Transportation Research Part F, 62, 349–375.
- Wang S., Jiang Z., Noland R.B., Mondschein A.S., 2020. Attitudes towards privately-owned and shared autonomous vehicles. Transportation Research Part F, 72, 297–306.
- Panagiotopoulos I., Dimitrakopoulos G., 2018. An empirical investigation on consumers' intentions towards autonomous driving. Transportation Research Part C, 95, 773–784.



## **Autonomous Driving & Travel Demand Modeling**





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A cross-sectional research that aims at **explaining individuals' behavioral intentions** w.r.t. different **potential approaches to autonomous driving**, investigating both **observable and latent factors**, and **profiling travelers' segments**.

#### **Research Questions**

- RQ1: What are individuals' perceptions, expectations, safety concerns, and intentions towards autonomous driving?
- **RQ2:** How will autonomous driving change people's travel behaviors?
- **RQ3:** Which factors most explain users' heterogeneity towards owning, sharing or riding autonomous vehicles?



## METHODOLOGICAL APPROACH





# I) DATA COLLECTION

## **Revealed Preference / Stated Intention (RP/SI) survey**

Observable factors

(Likert items indicators)

Latent factors

#### **Collection strategy**

- Revealed Preference (RP) / Stated Intention (SI) survey
- Computer Assisted Web Interviewing (CAWI)
- Random sampling
- 30 survey days (in January-February 2021)

#### Sections of the questionnaire

- Socio-economic characteristics
- Travel habits
- Perceptions about AVs
- Personal attitudes
- Intention-to-adopt AVs



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# II) LATENT VARIABLES IDENTIFICATION

#### **Measurement Models**

Latent variables can be inferred from some observed data, indicators, and differentiated through other called observable explanatory variables (age, gender, travel frequency, travel purpose, etc.). The indicators could be for example responses to attitudinal questions, perceptual and motivational surveys.

Latent variables are obtained through the specification of measurement models: essentially linear regression models where the main predictor, the factor, is latent or unobserved:

> $Y = \gamma Z + \tau$  $Z = \alpha' W + \omega$

where Y is a set of endogenous observed indicators (items), Z are the latent variables of interest, W is a set of observable exogenous multiple causes of Z,  $\alpha$  and  $\gamma$  are estimable parameters, and  $\omega$  and  $\tau$  are random errors.

#### e.g. MIMIC model in the case of one latent variable



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## III) MODEL ESTIMATION

### **Hybrid Choice Models: Ordered Logit**

Therefore in the case of a ordered choice model the latent utility function  $y^*$  can be written as:



where X are the attributes related to the transport solution and decision-maker, Z are the latent variables of interest,  $\beta$  and  $\delta$  are estimable parameters of the variables, and  $\varepsilon$  the random residual.

The continuous latent utility  $y^*$  is observed in discrete form through a censoring mechanism:

$y = 1$ if $-\infty < y^* < \mu_0$			$Prob(very unlikely) = Prob(y^* < \mu_0)$
	$= 2$ if $\mu_0 < y^* < \mu_1$		$Prob(unlikely) = Prob(\mu_0 < y^* < \mu_1)$
Observed counterpart of latent intention-to-use AVs	= 3 if $\mu_1 < y^* < \mu_2$	$\rightarrow$	$Prob(neutral) = Prob(\mu_1 < y^* < \mu_2)$
	= 4 if $\mu_2 < y^* < \mu_3$		$Prob(likely) = Prob(\mu_2 < y^* < \mu_3)$
	= 5 if $\mu_3 < y^* < +\infty$		$Prob(very likely) = Prob(y^* > \mu_3)$

Behavioral Models for Impact Assessment of Autonomous Driving on Travel Demand

## **RESEARCH RESULTS**

- RQ1: What are individuals' perceptions, expectations, safety concerns, and intentions towards autonomous driving?
  - Intentions to adopt autonomous vehicles are not directly correlated with the socio-economic characteristics of individuals
  - Perceived Usefulness is positively correlated with the willingness to adopt AVs, while Cost concerns is negatively correlated
  - Among the personal attitudes a slight correlation is observed between willingness to adopt AVs and technology-savviness (+ correlation), propensity for sharing (+) and aversion to Public Transport (-)
  - A strong positive correlation between the different willingnesses exists, testifying that those who are most willing to use AVs, on average, are so for all three potential approaches (own, share, ride)



Spearman's rank correlation matrix. Statistical significance: \*\*\* p-value<0.001, \*\* p-value<0.01, \* p-value<0.05

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## **RESEARCH RESULTS**

- **RQ2:** How will autonomous driving change people's travel behaviors?
  - Users are more interested in sharing than owning AVs Ο
  - Users are more interested in individual rather than collective use Ο









I will use PT services with autonomous minibuses



MODEL 1	Pane	Panel Ordered Logit Model with latent variables			
# observations	-	1218			
Restricted log likelihood:		-16254.9			
Final log likelihood:		-10287.6			
Rho-squared		0.367			
Akaike Information Criterion		20681.2			
	•				
Observable Variable		Coeff.	t-ratio		
Constant	5.71	*	1.71		
Ownership (Yes = $1$ , No = $0$ )	-1.50	) ***	-7.83		
Individual use (Yes = $1$ , No = $0$ )	0.63	**	2.04		
Latent Construct		Coeff.	t-ratio		
Perceived Usefulness	1.31	***	4.02		
Safety concerns	-9.0	7 *	-1.69		
Cost concerns		7 **	-2.19		
Technology-savviness		***	3.74		
Propensity for sharing		**	2.40		
Aversion to Public Transport	-1.20	) ***	-4.59		
Thresholds Parameter		Coeff. t-ratio			
Mu(01)	1.42	***	12.20		
Mu(02)	4.26	***	20.40		
Mu(03)		***	17.80		

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\*\*\*, \*\*, \* Significance at 1%, 5%, 10% level.

Behavioral Models for Impact Assessment of Autonomous Driving on Travel Demand



## **RESEARCH RESULTS**

MODEL 2	Panel Ordered Logit Model with latent variables			
# observations	1218			
Restricted log likelihood:	-16254.9			
Final log likelihood:	-10162.9			
Rho-squared	0.375			
Akaike Information Criterion	20447.9			

- **RQ3:** Which factors most explain users' heterogeneity towards owning, sharing or riding autonomous vehicles?
  - Safety concerns is the latent variable that impacts the most, followed by the cost concerns
  - Personal attitudes allow the profiling of the different demand segments
  - Some observable factors become significant when they are used to explain the heterogeneity in means of some latent traits

Observable Variable		eff.	t-ratio
Constant	3.60		1.48
Ownership (Yes = 1, $No = 0$ )	-1.74	***	-8.09
Individual use (Yes = $1$ , No = $0$ )	0.92	***	2.69
Gender (Female = 1, Male = $0$ )			
Age (More than 45 y.o. $= 1$ , otherwise $= 0$ )		*	-1.81
Education (PhD or Master's degree = 1, otherwise = $0$ )			
Income (More than $2.500 \in = 1$ , otherwise = 0)			
Household type (Live alone = 1, otherwise = $0$ )	-1.22	***	-3.69
City extension (More than $100.000$ inhabitants = 1, oth. = 0)			
Travel frequency (More than 4 times per week $= 1$ , oth. $= 0$ )			
Latent Construct	Coeff.		t-ratio
Perceived Usefulness	0.46	**	2.08
Safety concerns	-7.42	**	-2.21
Cost concerns	-5.96		-1.64
Technology-savviness	2.57	***	3.43
Propensity for sharing	1.10	***	3.40
Aversion to Public Transport	-1.37	***	-4.77
Heterogeneity in Mean	Coeff. t-ratio		t-ratio
Perceived Usefulness   Household type	-0.13		-1.50
Safety concerns   Gender		**	-1.97
Cost concerns   Income	-0.10		-0.65
Technology-savviness   Age	-0.54	***	-5.43
Propensity for sharing   City extension		***	4.44
Aversion to Public Transport   Preferred mode: Car	-0.67	***	-5.33
Thresholds Parameter		eff.	t-ratio
Mu(01)	0.88	***	16.40
Mu(02)	2.38	***	12.30
Mu(03)	5.36	***	20.70

\*\*\*, \*\*, \* Significance at 1%, 5%, 10% level.



# **CONCLUSIONS & RESEARCH PERSPECTIVES**

 The SP/SI survey designed with the Likert method allows to effectively measure the personal attitudes of individuals and their expectations on (any possible) emerging technology

 The behavioral models with latent variables allow to highlight the possible existence of heterogeneity among individuals

- Including latent traits of individuals in discrete choice models leads to the estimation of more robust models
- Observable factors (such as socio-economic characteristics and travel habits) can be explanatory exogenous variables of the heterogeneity in means of the latent variables
- When dealing with uncertainty, as in the case of a disruptive technology such as autonomous driving, the latent factors explain most of the travelers' intention-to-adopt a new transport solution



# **CONCLUSIONS & RESEARCH PERSPECTIVES**

- The consumption-based (as a service) approach to autonomous driving rather than ownershipbased (as a product) is predominant, but also an individual rather than collective use of AVs
- Age has been found to be a determinant of the intention to adopt AVs: this result bodes well given that the new generations will be the potential users of these transport solutions.
- Gender gaps in expectations about autonomous driving exist: females on average have a stronger sense of safety concern
- The costs for taking advantage of this new technology are an important issue

- Automation could be the driving force for a sustainable development of future mobility, which will allow to overcome the current car-ownership model even if not the car-oriented model of urban mobility
- Transport policies will need to be inclusive, and take into account the needs of vulnerable demand segments such as the elderly and females
- Travel Demand Management measures that encourage collective transport will foster social equity, meeting the needs of those with less economic resources
- Research now have to focus on the design and simulation of such policies for the evaluation of the wider impacts of autonomous driving on environment, society and land use





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# Thank you for your attention!

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