What We Know about the Use of Ridehailing: Frequency of Use and Impacts on Other Modes



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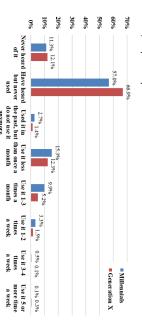
Latent Class Analysis of Behavioral Changes

Latent Class Analysis

Identification of different classes of behavioral changes

Frequency of Use of Ridehailing

probit model to control for sample selection and inflation in the number of ordered probit model with sample selection and (b) a zero-inflated ordered travel-related choices, and attitudes and perceptions. We estimate (a) an sociodemographic, built environment, use of technology and social media, analyze the impacts of five main groups of explanatory variables: and members of Generation X using the California Millennials Dataset. We zeros, respectively. The results are consistent across models ridehailing services, such as those offered by Uber and Lyft, among millennials We investigate the factors that affect the adoption and frequency of use



	Durdawad D		Thun infiate	L Durland
	Sample Selection	election	Probit Model	Model
	Selection	Frequency	Inflation	Frequency
	Estimates	Estimates	Estimates	Estimates
Age and Stage of Life	(r-vaives)	(F-Value)	(conine)	(Former d
Younger Dependent Millennials	0.22 (0.25)	1	0.28 (0.20)	;
Younger Independent Millennials	0.50 (0.00)	1	0.59 (0.00)	:
Older Dependent Millennials	0.32 (0.10)	;	0.29 (0.18)	:
Older Independent Millennials	0.56 (0.00)	1	0.59 (0.00)	:
Younger Gen X	0.21 (0.10)	;	0.23 (0.10)	;
Education				
High (Bachelor's degree or higher)	0.26 (0.00)	:	0.29 (0.00)	;
Presence of Children in the Household				
Household with Kid(s)	-0.28 (0.00)	;	-0.22 (0.02)	;
Region				
San Francisco Bay Area	0.08 (0.59)	:	0.12 (0.46)	:
Sacramento	0.20 (0.21)	:	0.18 (0.33)	:
Greater Los Angeles	0.22 (0.12)	:	0.29 (0.06)	;
San Diego	0.38 (0.01)	:	0.44 (0.01)	;
Built Environment				
8-Tier Employment Entropy	:	-0.45 (0.03)	:	-0.59 (0.06)
Standardized Activity density	:	0.18 (0.00)	:	0.22 (0.00)
Transit Performance Index	0.05 (0.00)	:	0.04 (0.02)	;
Use of Smartphone and Technology Adoption				
Use of Smartphone to Determine Destination and Route	0.21 (0.00)	0.18 (0.03)	0.20 (0.00)	0.36 (0.00)
Use of Other Emerging Transportation Services	1 01 00 001	0.0000		
Ereninger of Heing Tayl Canding	fooio) sois	1000	looid) acid	
Used less than Once a Month	0.35 (0.00)	:	0.50 (0.00)	:
Used at Least Once a Month	0.51 (0.00)	1.09 (0.00)	0.77 (0.00)	1.09 (0.00)
Frequency of Long Distance Travel				
Frequency of Non-car Long Distance Business Travel	0.13 (0.04)	1	1	1
Frequency of Long Distance Leisure Travel by Plane	0.43 (0.00)	0.17 (0.09)	0.50 (0.00)	0.32 (0.00)
Vehicles Per Household Driver				
Zero-Vehicle Household	;	0.89 (0.01)	;	0.69 (0.06)
Attitudes and Perceptions				
Variety Seeking	0.13 (0.01)	:	0.11 (0.03)	1
Technology Embracing	0.21 (0.00)	;	0.22 (0.00)	;
Pro-Environmental Policies	0.12 (0.00)	:	0.12 (0.01)	
Pay to Reduce Travel Time		1	-	0.18 (0.02)
Perceived Uber/Lyft Cost- & Time-related Limitations	-0.12 (0.02)	:	-0.12 (0.03)	:
Preference to Use Non-car Mode	-0.17 (0.00)	:	-0.19 (0.00)	;
Knowledge about the Services	-0.33 (0.00)	-	-0.36 (0.00)	-
Preterence to Use Own Vehicle	-0.13 (0.00)	-0.12 (0.04)	-0.11 (0.05)	-0.32 (0.00)
Correlation Parameter (p) Final Model Loglikelihood	-0.51 (0.00)	24	-787.86	- 26
			- 10/	

Main Findings from Frequency Models

Individuals from zero-vehicle households use Uber/Lyft more frequently Sociodemographics are better predictors of adoption than frequency

•

- Frequent long-distance travelers (by plane, in particular) use Uber/Lyft more often.
- significant in the adoption model Geographic region and public transit quality and connectivity are only
- Land-use mix and population + job density impact the frequency of use

 Least affluent Cost and time sensitiv

Urban dwellers

with good transit acce

with poor transit access

Older Gen Xers

urban area

accessibility

Multimodal travelers Frequent commuters Younger/indepe Millennials

Like biking

Multimodal when possible Non-frequent commuters Want to come back to Not cost and time sensit Low transit and walk Suburban Dwellen

Pro-environ

Class 3 (size=10%)

Most frequent users c

Uber/Lyft

- trequent users. Those that prefer to own/use their own vehicle are less likely to be of ridehailing.
- imes The higher the frequency for *carsharing* , the lower the frequency for Competition with other shared-mobility services: Uber/Lytt.

Potential Impact of Ridehailing



59.5%

Class 2 (size=37%)

Class 3 (size=10%)

93.6%

- We asked individuals to report how the use of ridehailing impacts the use
- of other modes (based on their last trip made by Uber/Lyft).

Pooling is the answer!

externalities.

factors to the use of ridehailing

Cost and preference toward the use of personal vehicle are limiting

Less Drive Less Walk/Bike Less Transit

More Walk/Bike

More Transit

Class 3 0.0% 3.0%

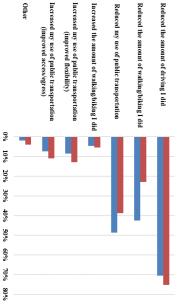
Class 1

2.3% 0.0%

0.6% 1.7% 0.5% 0.1%

21.0%

- Multiple answers were allowed for each respondents.
- We performed latent class analysis (LCA).
- Three rather well defined latent classes were identified in our preliminary
- Next step is to control for individual differences using active covariates



better service, with lower economic and environmental costs.

Shared mobility can be integrated with public transit to provide Opportunities for demand-responsive services and microtransit.

Single-passenger ridehailing tends to (a) substitute for driving, (b) groups), and (c) increase the attractiveness of living without a car:

It is a case where the public interest aligns with business interests.

Policymakers need better understanding of who might use pooling

services and what incentives/policies could encourage its use.

Pooling is the primary strategy to reduce prices and negative

replace the use of transit or active modes (especially among some

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Generation X Millennials their help with survey design, data collection and cleaning.