

Introduction

- Usage-based motor insurance (UBI) schemes, are a new innovative concept that has commercialized around the world.
- Pay-as-you-drive (PAYD) and Pay-how-you-drive (PHYD) are two very popular schemes.
- Recent schemes but with very promising practice.
- Significant potential impact on traffic safety, traffic congestion mitigation and pollution emiss

Objectives

- The main objective is to investigate which parameters affect users' willingness to pay based motor insurance pricing schemes.
- Pay-as-you-drive (PAYD) and Pay-as-how-you-drive (PHYD) schemes are chosen.
- A dedicated questionnaire was designed and administered to participants.
- A mixed logit model is implemented to investigate effect of driving characteristics, driver price of vehicle insurance premiums on vehicle insurance choice.
- The findings of the study are expected to extend previous research and add to current kno

Data

- A dedicated questionnaire designed including both revealed preference questions ab insurance type as well as stated preference scenarios related to current and alternative ins
- \checkmark To increase the number of alternative tested scenarios, two different sheets were design eight scenarios PHYD each and each of the 100 respondents answered a single shee structured in 4 sections and questions included:
- general respondent's driving data (years since license was obtained, vehicle make, curren driving behavior data
- alternative stated preference scenarios about the new insurance premium policies (PAY benefits
- personal demographic data.
- The required time for completion was 10-12 minutes and it was administered to drive motorist's service station in Attica region in Greece.
- As for the number of scenarios chosen, it was decided that for the proper implementation number of scenarios should be reduced.
- Sector Based on the number of possible values that the variables of the stated preference question take, the number of different scenarios results to 16 for PAYD and 80 for PHYD.
- The number of different combinations in this study was reduced based on an orthogonal implemented, under the assumption that no correlations between typical alternatives exist.
- ✓ In stated preference surveys fractional factorial design can be used instead of full factorial
- Both these designs ensure orthogonality however, the full factorial design would include respectively, in contrast to the fractional comprising fewer combinations and are gua desirable statistical properties such as the identification and accuracy.

Method of Analysis

- The core analysis of this study is the mixed logit model (random parameters logit mode
- Superior to the fixed effects model
- The mixed logit model is used to account for potential unobserved heterogeneity.
- Assumes that the estimated parameters vary across observations.
- Each variable set as random follows a distribution (e.g. normal, uniform)
- Widely used in discrete choice experiments.
- Estimation of the mixed logit model takes place by using simulation methods due to the difficulty in computing probabilities.
- A mixed logit model is any model whose choice probabilities can be expressed in the form: $P_{ni} = \int L_{ni}(\beta) f(\beta) d\beta$

• where $Lni(\beta)$ is the logit probability evaluated at parameters β : $\rho^{V}ni(\beta)$

$$L_{ni}(\beta) = \frac{e^{V_{nj}(\beta)}}{\sum_{j=1}^{J} e^{V_{nj}(\beta)}}$$

 Then the mixed logit probability takes the usual form: $P_{ni} = \int (\frac{e^{\beta' x_{ni}}}{\sum_{i} e^{\beta' x_{nj}}}) f(\beta) d\beta$

Willingness-to-Pay for Usage-Based Motor Insurance Athanasios Theofilatos, Dimitrios I. Tselentis, George Yannis, Manos Konstantinopoulos

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recently started to be												
	Descriptive statistic	S										
	INDIVIDUAL SPECIFIC						ALTERNATIVE SPECIFIC					
	VARIABLES		Mean	St.deviation		Max.		Abbreviatio	n Mean	St.deviation	Min.	N
ons reduction.	Gender = Female Age: 18-25 (reference	GENDER_F	0.45	0.50	0.00	1.00	PRESENT INSURANCE % change in mileage					
	category)	AGE1	0.04	0.20	0.00	1.00	(current Insurance)	KM	0.00	0.00	0.00	(
	Age: 25-30	AGE2	0.07	0.26	0.00	1.00	% change in Insurance		0.00	0.00	0.00	•
	Age: 3040	AGE3	0.43	0.50	0.00	1.00	Cost (current Insurance)	COST	0.00	0.00	0.00	
for alternative usage-	Age: 40-50	AGE4	0.28	0.45	0.00	1.00	% change in Speed					
	Age: >50	AGE5	0.11	0.31	0.00	1.00	(current Insurance)	SPEED	0.00	0.00	0.00	
	PC usage is made	USAGE_PC	0.98	0.14	0.00	1.00						
	Smartphone Owner	SMARTPHONE	0.78	0.41	0.00	1.00	PAYD INSURANCE					
	Married	MARRIED	0.53	0.50	0.00	1.00	% change in mileage					
	Income <10000 (reference category)	INCOME1	0.06	0.24	0.00	1.00	(PAYD Insurance)	KM	-11.76	6.58	-20.00	
	10000 < Income < 25000	INCOME2	0.54	0.24	0.00	1.00	% change in Insurance	000T	44.00	0.00	00.00	
emographics, and the	Income > 25000	INCOME3	0.40	0.49	0.00	1.00	Cost (PAYD Insurance)	COST	-11.69	6.63	-20.00	
	Occupation: Public Sector	OCCU1	0.45	0.50	0.00	1.00	PHYD INSURANCE					
dao	Occupation: Private Sector	OCCU2	0.24	0.43	0.00	1.00	% change in mileage					
edge.	Occupation: University						(PHYD Insurance)	КM	-6 25	9 61	-20	
	Student	OCCU3	0.03	0.17	0.00	1.00	% change in Insurance	KM -6.25 9.61 ce ce) COST -11.43 6.78	20			
	Occupation: Freelancer	OCCU4	0.09	0.29	0.00	1.00	Cost (PHYD Insurance)	COST	-11.43	6.78	-20.00	
	Occupation: Enterpreneur	OCCU5	0.03	0.17	0.00	1.00	% change in Speed (PHY					
t current vehicle and	Occupation: Household		0.02	0.14	0.00	1.00	Insurance)		-11.47	6.80	-20.00	
ance schemes.	Occupation: Technician Occupation: Pensioner	OCCU7	0.00	0.00	0.00	0.00	i					
	(reference category)	OCCU8	0.07	0.26	0.00	1.00						
d with four PAYD and	Occupation: Unemployed	OCCU9	0.02	0.20	0.00	1.00						
t. The questionnaire is	Occupation: Other	OCCU10	0.05	0.22	0.00	1.00						
	Education: Primary											
nsurance cost etc.)	Education	EDU1	0.03	0.17	0.00	1.00						
	Education: Secondary											2
and PHYD) and their	Education (reference											
	category)	EDU2	0.24	0.43	0.00	1.00			m	1		
	Education:Technological		0.04	0 47	0.00	1 00		1	100			
re boing stopped at a	Educational Institute Education: University	EDU3	0.34	0.17	0.00	1.00			3		-	
s being stopped at a	Degree	EDU4	0.11	0.31	0.00	1.00						
	Education: Postgraduate		0	0.01	0.00	1.00						
	Degree	EDU5	0.24	0.43	0.00	1.00				1		
on of the research the			0.00	0.17							-	
n of the research the	Education: Ph.D.	EDU6	0.03		0.00	1.00						
on of the research the			0.03	0.17	0.00 0.00	1.00 1.00						
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naire were designed to	Education: Ph.D. Education: Other PAYD model Variables	EDU6 EDU7	0.03	0.17		1.00	Standard error p-	-value (Conclusi	ion C	Odds 1	ra
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variables (constant term and cost) are considered as fixed.

• According to the Z score table and the normal distribution function, some 3.52% of observations are lower than zero. This means that in about 96.48% of observations, Km is associated with increased likelihood of selecting PAYD while only 3.52% of observations show a negative correlation. Therefore, as offered percentage reduction in driven mileage decreases, it is more likely that the driver chooses the PAYD policy.

• The cost parameters was considered as fixed, therefore, the negative sign of the beta coefficient (-0.154) denotes that as the cost reduction is lower, drivers are more likely to choose the present insurance.

People with primary education are more likely to choose PAYD.

• The negative value of the coefficient of USAGE_PC variable (-3.93), denotes that drivers who are more familiar with personal computer usage are more likely to choose the present insurance rather than the PAYD policy.

• Familiarity with smartphone use is more likely to make drivers choose the PAYD policy.

• The gender variable shows that female drivers tend to prefer the PAYD compared to males. More specifically, probability to select PAYD is almost twice higher than males.



Results (2/2)

PHYD model

Variables **Random parameters (normal** Constant term Standard deviation of constant te Km Standard deviation of Km Cost Standard deviation of Cost Speed Standard deviation of Speed **Fixed parameters** AGE4 AGE5 **SMARTPHONE** GENDER_F

Log-likelihood of the empty i Log-likelihood of the full mod McFadden's pseudo R²

- PHYD.
- PHYD.
- respectively
- (similarly to the PAYD) compared to the present policy.

Conclusions - Discussion

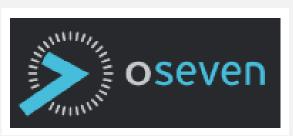
 A methodological approach is proposed to identify the parameters that affect users' willingness to pay for alternative usage-based motor insurance pricing schemes such as PAYD and PHYD.

 Kilometers and cost reduction were also found to affect similarly the choice for both UBIs i.e. the higher the kilometers reduction the lower the probability of the UBI scheme to be chosen and the higher the cost reduction the higher the probability of the UBI scheme to be chosen by a user. Moreover, the higher the speed reduction imposed to the user the lower the probability of the UBI scheme to choose it.

- Future research could carry out surveys: \geq on a national sample > in different countries
- different scenarios
- \succ including more parameters.
- ranking of alternatives schemes
- best-worst scaling modeling

Acknowledgement

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	Estimate	Standard error	p-value	Conclusion	Odds ratio
l distribution)					
	-1.789	0.429	0.000	95% significant	0.167
term	1.197	0.270	0.000	95% significant	-
	0.114	0.017	0.000	95% significant	1.121
	0.061	0.027	0.022	95% significant	-
	-0.179	0.025	0.000	95% significant	0.836
	0.065	0.025	0.009	95% significant	-
	0.091	0.020	0.000	95% significant	1.095
	0.077	0.022	0.001	95% significant	-
	-0.846	0.274	0.002	95% significant	0.429
	-1.176	0.433	0.007	95% significant	0.309
	0.627	0.309	0.042	95% significant	1.872
	1.005	0.244	0.000	95% significant	2.731
model	513.250				
del	-416.500				
	0.216				

• The variables "Km", "Speed", "Cost" and the constant term, were set as random following the normal distribution. Km has a mean value of 0.114 and a standard deviation of 0.061, Cost has a mean of -0.179 and standard deviation 0.065, while Speed has a mean value of 0.091 and 0.077. On the other hand, the constant term was found to have a mean value of -1.789 and standard deviation 1.197.

• Concerning Km, the calculated Z-values indicate that 97% of observations have a positive correlation with

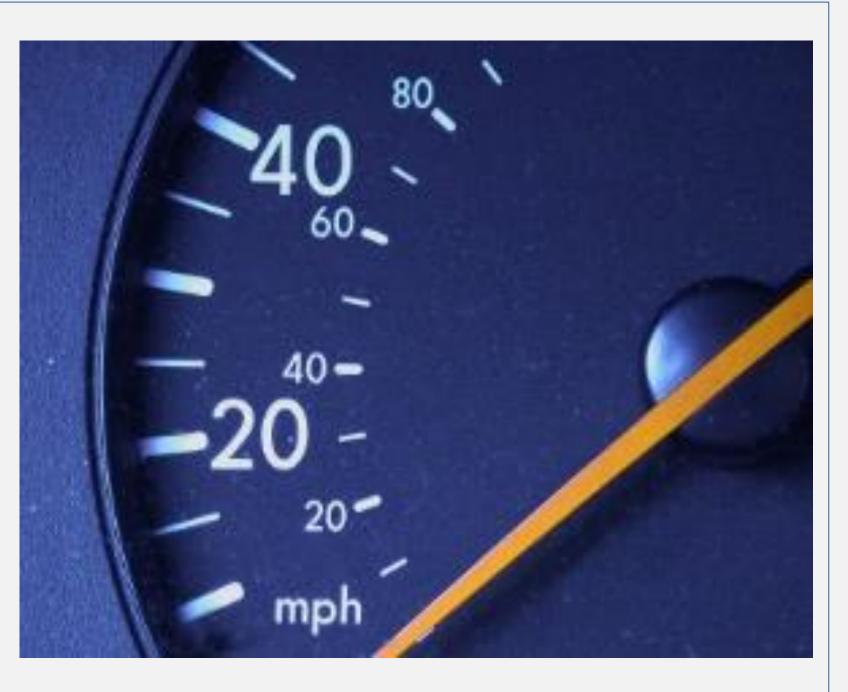
• Regarding speed, indicate that about 11% of observations have a negative association with PHYD while 89% have a positive association with PHYD. This means that as the percentage reduction in speed tends to zero, the driver is more likely to choose the PHYD policy scheme.

• The variable Cost has a negative mean value as in the previous model, indicating that the percentage reduction in cost tends to be zero, the present policy is more probable to be selected by drivers than the

• Drivers 40-50 years old and older than 50 years old are more likely to prefer the present insurance policy compared with younger drivers. More specifically, young drivers are almost 2.5 times and almost 3 times more probable to choose the PHYD policy, compared to drivers 40-50 years old and older than 50 years old

• Familiarity with smartphone and applications suggests high probability for drivers choose the PHYD scheme

• The gender variable shows that female drivers would prefer the PHYD compared to male drivers.



> alternative models to account for heterogeneity could be utilized, for example the latent class model.