

Extension of an adoption model to evaluate autonomous vehicles acceptance

Iman Farzin¹, Amir Reza Mamdoohi^{2*}, Francesco Ciari³

¹ Ph.D., graduated, Faculty of Civil and Environmental Engineering, Tarbiat Modares University, Tehran, Iran, P.O.Box: 14115-111, Mobile: +989127848553, Email: Iman.farzin@modares.ac.ir

² Associate professor, Faculty of Civil and Environmental Engineering, Tarbiat Modares University, Tehran, Iran, P.O.Box: 14115-111, Phone: +982182884925, Mobile: +989123358497, Email: Armamdoohi@modares.ac.ir

³ Associate Professor, Department of Civil, Geological and Mining Engineering, Polytechnique Montréal University, Montréal, Canada, P.O.Box: 6079, Phone: +15133404711, Email: Francesco.ciari@polymtl.ca

Abstract

Autonomous Vehicles (AVs) can provide safe, clean and efficient mobility by using advanced communication technologies to create an unprecedented revolution in transportation. Acceptance of AVs has a key role in their successful implementation. Most researchers have used Technology Acceptance Model (TAM), Theory of Planned Behavior (TPB) and Unified Theory of Acceptance and Use of Technology (UTAUT) to identify latent factors affecting, which focus only on individuals' internal schema of beliefs without considering the external factors of acceptance. The current study, uses Trialability (TR), Observability (OB) extracted from Diffusion of Innovations (DOI) theory, Performance Expectancy (PE), Effort Expectancy (EE), Social Influence (SI) extracted from UTAUT, as well as Perceived Risk (PR), Environmental Concerns (EC) and Consumer Innovativeness (CI) to identify a wider set of latent factors. A stated preference survey conducted to this purpose in Tehran allowed collecting 641 responses. Considering the latent nature of research variables, Structural Equation Modeling is applied. Results show that PE, EE, PR, OB, SI, TR, CI and EC affect acceptance in decreasing order of regression weights, an explain 72.5% of the variance in the dependent variable.

Keywords: Acceptance of Autonomous Vehicles, Trialability, Observability, Performance Expectancy, Effort Expectancy, Social Influence, Perceived Risk, Environmental Concerns, Consumer Innovativeness, Structural Equation Modeling

1- Introduction

Recent technology developments have reduced the scope of human intervention in vehicle movement in the transportation network. They will finally lead to the emergence of fully-automated autonomous mobile robots. These robots move without any human interventions and can change the future of transportation [1]. They are vehicles with integrated multi-sensor navigation and intelligent decision making systems [2] that improve the road safety (by reduce/eliminate human error) [3], increase network capacity, improve traffic flow efficiency, use the available capacity optimally [4], improve fossil fuel consumption [5], enrich travel time [6], improve land use patterns [7], reduce costs, increase social wealth [6], increase urban access [8] and, finally, achieve sustainable [9] and intelligent transportation goals [10].

Several studies, however, are concerned about the possibility of negative consequences from these vehicles, including an increase in reliance on private cars [11], a reduction in the share of public transportation [12], and the affordability of purchasing such vehicles based on their expected high price [9].

37 In the early stages of the introduction of any new product, policymakers should be aware of the
38 needs and factors that affect people decision whether to accept and use it or not [13]. Knowing
39 the potential users' perspective can help a further growth of the technology, facilitates its
40 implementation and lead to a better evaluation and prediction of the users' responses and, hence,
41 results in an optimal design and effective future development and planning and development
42 [14]. Often, emerging technologies face unpopularity in their early market-introduction stages
43 because consumers follow an "initial perception-resistance-gradual adaptation-final absorption"
44 cycle, and technologies resisted hard at the beginning become essential to people's lives over
45 time [15]; the same is quite likely to happen with AVs [16].

46 Among different AVs classifications proposed by different authorities, a very popular one is
47 presented by Society of Automotive Engineers (SAE) and has six different automation levels: 1)
48 no automation (level zero), 2) driver assistance, 3) partial automation, 4) conditional automation
49 (autonomous in special traffic), 5) high automation (autonomous on specific infrastructure) and
50 6) full automation [17]. This study examines the last one.

51 There are many factors that contribute to acceptance of technology. These factors can be divided
52 into internal and external categories. Internal factors (such as Performance Expectancy (PE),
53 Effort Expectancy (EE), Social Influence (SI), Perceived Risk (PR), and Consumer
54 Innovativeness (CI)) originate solely from people's attitudes towards technology. While external
55 factors (such as Trialability (TR), Observability (OB), Environmental Concerns (EC) are
56 influenced by a combination technology availability, the surrounding environment and
57 individual's attitude. According to the literature review, the effect of each of these internal and
58 external factors on technology acceptance can be decreasing or increasing.

59 Most of the previous studies have used the Technology Acceptance Model (TAM), Theory of
60 Planned Behavior (TPB) and Unified Theory of Acceptance and Use of Technology (UTAUT) to
61 identify the latent factors of the acceptance of the AV technology [16]. Considering that human
62 behavior is complex, focusing on only one group of variables (internal or external categories)
63 could lead to incomplete results. Motivated by addressing the aforementioned gaps, the main
64 contributions of this paper are:

65 To the best of the authors' knowledge, a few studies considered the combination of internal or
66 external factors [17,18]. We aim to integrate internal and external factors proposing a more
67 comprehensive model that overcomes the limitations of previous studies in AV acceptance. We
68 combine TR, OB, EC as external factors, with PE, EE, SI, PR and CI variables as internal factors
69 (Fig . 1).

70 The majority of studies on acceptance of AVs have been conducted in developed countries [19-
71 26]. Research indicates that common beliefs and values of a society influence the acceptance of
72 technology. Moreover, the impacts of values on technology acceptance vary in different
73 countries, which means that the results of research conducted in developed countries should not
74 be generalized. The current study employs data collected in Tehran, Iran, which is a developing

75 country. Our work, therefore, is a contribution because there is little literature about the issue in
76 developing countries.

77 The literature review (theories, latent variables) is in Section 2, we explain the questionnaire and
78 analyze the statistical data in Section 3, we present the modeling results in Section 4 and provide
79 conclusions and suggestions for future studies in Section 5.

80 **2- Literature review**

81 The two general classes of previous studies on the latent factors affecting the acceptance of AVs
82 are 1) those using the behavioral theories and 2) those using variables other than those used in
83 class 1 [27]. Among different theories developed to describe the technology acceptance and its
84 affecting factors, the Theory of Reasoned Action (TRA), Theory of Interpersonal Behavior
85 (TIB), Social Cognitive Theory (SCT), TPB, DOI, TAM, Motivational Model (MM), Uses and
86 Gratification Theory (UGT), Model of PC Utilization (MPCU) and UTAUT are worth
87 mentioning [28]. This study uses the DOI and UTAUT as the underlying theories. In the
88 following sections, related explanations are given to clarify and justify the reason behind this
89 selection. Also, other related factors namely PR, EC, and CI are discussed.

90 **2-1- Diffusion of Innovations (DOI)**

91 The DOI theory identifies, by quantitative tools, the diffusion rate of an innovation and factors
92 influencing its acceptance/non-acceptance to facilitate its implementation. In this theory, the
93 process of deciding to accept/reject an innovation is mental and people is assumed to pass from
94 the awareness to the acceptance/rejection and finally to the confirmation stage through five steps:
95 1) knowledge, 2) persuasion, 3) decision-making, 4) Implementation, and 5) confirmation.

96 In Step 1, people gain information on an innovation and learn how it works.

97 In Step 2, people develop a favorable/unfavorable attitude towards the innovation. Factors
98 playing important roles in creating a positive attitude are:

99 a) Relative advantage - an individual's belief that the new innovation tops the previous ones (the
100 main issue, as Rogers' theory states, is how people see an innovation and if it is really
101 beneficial).

102 b) Compatibility-a person's belief if the new innovation is in harmony with the existing values
103 and his/her past experiences and needs. If the answer is negative, the acceptance rate will
104 decrease.

105 c) Complexity - difficulty in using the innovation perceived by the individual.

106 d) Trialability - addresses the innovation's reviewability and testability; the pre-belief that an
107 innovation can be tried and experienced will affect its acceptance/rejection.

108 e) Observability – individuals' seeing/feeling the innovation results; if so, as Rogers believes, it
109 is more likely to be accepted by users.

110 In Step 3, people decide to totally accept an innovation at the beginning, totally reject it at the
111 beginning or accept it open-mindedly at first with the option to reject it after a while.

112 In Step 4, those who accepted the innovation try to use it. Here, people are still looking for
113 information and may change their mind if they hear conflicting messages.

114 In Step 5, those who accepted the innovation seek to justify their decision by emphasizing its
115 usefulness and benefits [29, 30].

116 The current research uses both the TR and OB variables in its conceptual model and suggests the
117 following hypotheses based on the results of previous studies [31-35]:

118 *H1: TR positively influences the acceptance of AVs.*

119 *H2: OB positively influences the acceptance of AVs.*

120 **2-2- Unified Theory of Acceptance and Use of Technology (UTAUT)**

121 UTAUT was formulated by assessing similarities and differences across eight models (including
122 TRA, TAM, MM, TPB, MPCU, DOI, SCT, TAM-TPB combination). The final significant
123 constructs were PE, EE and SI, and Facilitating Conditions, among which the first three affect
124 the behavioral intention, and the fourth affects the user behavior.

125 PE shows one's view to use technology to improve performance and EE relates to its easy use,
126 application and social impacts based on how important people think about his/her use of that
127 technology [36]. Previous studies have directly proven the effects of these latent factors on the
128 AV acceptance [37-39]. Following hypotheses are proposed:

129 *H3: PE positively influences acceptance of AV.*

130 *H4: EE positively influences acceptance of AV.*

131 *H5: SI positively influences acceptance of AV.*

132 **2-3- Non-Behavioral Factors**

133 Perceived Risk (PR) is defined as the occurrence of a probable loss [40] and loss and uncertainty
134 are its two main aspects. It plays an important role in a person's willingness to buy (use) a new
135 product and is increased with an increase in the damage expectancy [41; 42]. [43] claim that the
136 PR includes performance, financial, time, safety, social, and psychological risks, but [44] state
137 that it involves social, financial, physical, performance, time and psychological risks. Some
138 studies have proven that this variable has negative effects on the acceptance of AVs [45] and
139 other have shown that although it does not directly affect the willingness to use such vehicles, it
140 affects the people's trust level indirectly [32]. The present study defines this variable as the
141 potential risk of the technology for users to achieve the desired results (safe journey) and
142 proposes the following:

143 *H6: PR negatively influences acceptance of AV.*

144 People's EC include their considerations, interests/disinterest [46] and awareness of the
145 environmental risks [34] and also, its related emotional involvements [47]. Regarding the AV
146 impacts on the environment, while some believe they help protect the environment by affecting
147 factors such as speed, economical driving, reduced congestion, vehicle weight, moving in a
148 single lane, and reduced accidents [48]. This study proposes the following:

149 *H7: EC positively influences acceptance of AV.*

150 CI refers to people's different responses to new products and ideas. It leads the person to accept a
151 new product regardless of its price and quality [49]. It is also defined as a person's degree of
152 adaptation to a new product sooner than others and is studied as a force that leads to a novelty-
153 seeking behavior which can also be defined as a person's speed of acceptance of a new product
154 or his/her curious behavior to obtain information about it [50; 51]. Effects of this latent variable
155 on the acceptance of an AV show that people with such attitudes are more inclined to accept it
156 because it uses the latest technology [52; 17]. This study proposes the following hypothesis is:

157 *H8: CI positively influences acceptance of AV.*

158 This study uses data collected in Tehran. Fig. 2 shows the conceptual model of this study.

159 **3- Data**

To verify the proposed model, we conducted a face to face survey and collected data from Tehran (Iran's capital and largest city). According to some previous studies, AVs offer a variety of substantial benefits that are expected to revolutionize the transportation industry in the future such as increasing traffic flow efficiency [43, 44], allowing optimal use of transport infrastructures [45], so these technology can address some of the transportation problems in Tehran.

160 The pilot study was conducted in the spring of 2019 in the main parks of Tehran. In this study,
161 100 stated preference questionnaires were given to the interviewers and they were asked to write
162 down the items that the respondents are unclear about. The interviewers were also asked to
163 record the entire questioning process. The researchers reviewed all the recorded cases and made
164 changes in the questionnaire based on them as well as the opinions of the interviewers. After
165 these changes, the main study was conducted between July and September 2019 in cinemas,
166 parks and main squares of Tehran under the full supervision of researchers (figure (3)). The
167 revised questionnaire was randomly distributed among 22 Districts of Tehran.

- 168 • Individuals are first informed of the study objectives and the information confidentiality
169 and then shown a short clip to get acquainted with AVs and how to use them to meet their
170 transportation needs. In the clip, effort is made to provide enough information about the
171 technology without directing their responses.
- 172 • Next, items related to the used latent variables are extracted from different references and
173 individuals are asked, in the second part of the questionnaire, to respond the questions on
174 a 5-point Likert scale (from strongly disagree (1) to strongly agree (5)). The literature
175 suggests that five-point scale appears to be less confusing and to increase response rate.
- 176 • The last part is devoted to questions related to individuals' demographic characteristics
177 (gender, marital status, age, education and family size).

178 **3-1- Data analysis**

179 Different opinions have been expressed in order to determine the sample size in structural
180 equations modelling. Some believe that the ratio of the number of observations to the
181 independent variables should not be less than 5 [53]. Others have suggested a more conservative
182 10 ratio [54]. Based on the ratio of 8 for the number of observations to independent variables, the

183 desired number of samples is 560. For more assurance after refining and checking the outlier &
184 missing data, 641 valid sample are used. Table 1 presents the individuals' demographic profile.
185 The sample replicates Tehran's population distribution as of 54.4% men (collected data) vs.
186 52.3% (2016 census). The statistical analyses show that men, singles, aged 26-44, university
187 graduate and 4-member family size have the highest frequency among respondents.

188 This questionnaire contains 12 acceptance-related items and people are asked to respond such
189 questions as "I will buy if it is reasonably priced" to "I would recommend others to use/buy it"
190 according to five-point Likert scale of agreement (from 1= strongly disagree to 5= strongly
191 agree). They are also asked to respond such questions as "I will use it on optional trips
192 (shopping, leisure)" to "I will send it to store to buy the daily necessities" according to five-level
193 Likert scale of frequency (from 1= never to 5= always). Results show, in all questions except " I
194 will buy when the first model is released", that the number of people who said "agree" and
195 "strongly agree" (or "always" and "often") is more than those who said "disagree" and "strongly
196 disagree" (or "rarely" and "never"). The highest and lowest percent oppositions are for questions
197 of "I will buy when the first model is released" and "I will use it if it is reasonably priced",
198 respectively. People agree most with "I will use it if it is reasonably priced", and least with "I
199 will buy when the first model is released" (Fig. 4).

200 **4- Modeling results and discussion**

201 Fig. 5 and Table 2 show the SEM results achieved by the maximum likelihood method and
202 AMOS (Analysis of moment structures) 25 software. Due to the assumed relationships between
203 latent and observed variables (measurement models) as well as the assumed dependencies
204 between the various latent variables (structural model), we use SEM. This model is obtained
205 after several modeling runs, eliminating insignificant items or those with less standard regression
206 weights than an acceptable value to satisfy the evaluation criteria. Results show:

207 TR has positive and significant effects on the AV acceptance and conveys the concept that
208 technology can be reviewed and tested at a limited level to evaluate its benefits and usefulness.
209 Providing this possibility can lead to the innovation's more and sooner acceptance, and will
210 allow designers to detect and modify its weaknesses [30]. Some researchers emphasize the
211 importance of TR in accepting an innovation, especially in developing countries, because
212 facilities are inadequate and people are not sure if innovations adapt to the existing
213 infrastructures [55]. Using this variable can evoke people's emotions to accept an innovation and
214 it is suggested that the AVs' TR conditions be provided to turn as many potential users as
215 possible to actual ones.

216 OB is defined as the "degree of apparentness/tangibility of the innovation results" and has shown
217 to have significant and positive effects on the acceptance of autonomous cars because it
218 eliminates people's uncertainty and skepticism in using the technology [56]. If the AV
219 technology spread in the society and its benefits are made observable, its acceptance rate will
220 increase too.

221 Since PE has positive effects on the AV acceptance, any improvement in its efficiency will
222 strengthen the desire and, hence, the willingness to accept it. This result, conforming well to
223 those of other previous studies [37-39], highlights the importance of improving the performance
224 of these vehicles, especially in helping to achieve the transportation needs in an efficient and
225 effective way.

226 Results show that if a person can easily understand how to use an autonomous car and finds the
227 related skills, he will be more inclined to accept it. This means that the system should be so
228 designed as to allow the user to learn to use it more easily without needing much time and mental
229 effort because design is a very effective factor in accepting a new technology. Other researchers
230 too have had results similar to those of this study acknowledging the direct effect of this variable
231 on the AV acceptance [37-39].

232 The SI variable shows how much a person's feelings are affected by what his/her close and
233 important individual(s) think or suggest about his/her using a new technology [36]. Since
234 modeling results indicate that SI affects acceptance, the positive experience of one who has used
235 this car can affect his/her peers, impressionable coequals and, in general, those for whom he/she
236 is important; this conforms well to those of other studies on the AV acceptance [37-39].

237 An increase in the PR attitude reduces the AV acceptance which conforms well to the results of
238 other previous studies [57; 58]. Researchers have defined the PR as the consumer's perception of
239 the uncertainty and adverse consequences (if occur) of buying/using a product/service [59].
240 Since the PR can be reduced by increasing confidence and/or reducing consequences, it is
241 suggested that the AV designers should not only improve the car performance and minimize its
242 accident probability, but also make it so safe that the vehicle/passengers may experience the least
243 damage in case of an accident. With proper advertising, we can try to create the right mentality
244 about safety.

245 EC are the results of how one assesses the effects of one's behavior on the environment [60].
246 Similar to other studies [5; 60], this research finds that the effect of this variable on the AV
247 acceptance is significant and positive, which means people with more EC accept autonomous
248 cars more. Therefore, optimal routing, using clean fuels instead of the fossil type, reducing the
249 weight and sharing the use, thus, reducing the fuel consumption can help these cars be accepted
250 in the society more and more.

251 People with innovative attitude tend to use/buy new products faster than others because this
252 variable is a behavioral stimulus that drives a person to start and implement new ideas,
253 processes, and products [61]. Results of this study, consistent with those of other researches [57],
254 indicate that people with more innovative behavior accept AVs more because they see them as a
255 symbol of their desire due to the latest technologies used in such cars.

256 In the case, Likert scales are utilised for a study; Cronbach's alphas are considered the most
257 appropriate measures of reliability [58]. As shown in Table 3, Cronbach's alphas range from
258 0.722 to 0.906; thus, the constructs are deemed to have adequate reliability. Table 2 presents
259 standard regression weights for all items. Items loaded above 0.50 are considered for further
260 analysis. Therefore, both reliability and discriminant validity met the baseline criteria.

261 There are several criteria to evaluate the modeling of structural equations among which Chi-
262 squared/DOF < 5, goodness of fit index (GFI) > 0.9 and adjusted goodness of fit (AGFI) > 0.8 are
263 three indices [59]. The current research yielded CMIN/DOF = 3.245, GFI = 0.950 and AGFI =
264 0.839 for its proposed model which are acceptable, and RMSEA = 0.059 < 0.08. To evaluate the
265 model relative position between the worst and best fits, relative fit indices (incremental fit index
266 (IFI) and comparative fit index (CFI)) are recommended to be greater than 0.9 [60]; their values
267 in this study are 0.910 and 0.909, respectively (Fig. 5). Reliability and convergent validity
268 assessments using: 1) significant standard regression weights > 0.5, 2) Construct Reliability (CR)

269 > Average Variance Extract (AVE), 3) AVE> 0.5 and CR> 0.7 [58] show that the measurement
270 model fits well with the collected data (Table 3).

271 **5- Conclusions**

272 Autonomous vehicles (AVs) have the potential to fundamentally change the driver-vehicle
273 interactions and provide opportunities to dramatically improve the transportation efficiency,
274 stability and safety. This technology can reduce the fuel consumption by affecting such factors as
275 reducing congestion, routing optimally, less maneuvering, platooning and reducing accidents.
276 Further development of this technology to enable the best use of its features is tied to its
277 acceptance by the people. Careful analysis/studying of people's main reasons for
278 accepting/rejecting is of special importance to both decision makers and designers. Most
279 previous studies have used Technology Acceptance Model (TAM), Theory of Planned Behavior
280 (TPB) and Unified Theory of Acceptance and Use of Technology (UTAUT) theories to identify
281 latent factors that affect the AV acceptance. These theories have limitations because they
282 consider only the effects of individuals' internal schema of beliefs on acceptance. They neglect
283 the facilitating/hindering role of the external factors such as Trialability (TR), Observability
284 (OB) and Environmental Concerns (EC). As the aim of this paper, a more comprehensive model
285 considering the internal and external factors proposed to overcomes the limitations of previous
286 studies. Hence, this study used TR, OB (extracted from the Diffusion of Innovations (DOI)),
287 Performance Expectancy (PE), Effort Expectancy (EE), Social Influence (SI) (from UTAUT) as
288 well as Perceived Risk (PR), Environmental Concerns (EC) and Consumer Innovativeness (CI)
289 to identify the latent factors affecting the AV acceptance. Most studies were conducted in the
290 developed countries. To calibrate the proposed model, this research used the structural equation
291 modeling and data of 641 questionnaires collected randomly in Tehran (capital of Iran as a
292 developing country). Results of the statistical analyses of the responses to questions related to the
293 AV acceptance indicated that the number of people who chose "agree" and "strongly agree" (or
294 "always" and "often") in all items except "I will buy when the first model is released" was more
295 than the number of those who chose "disagree" and "strongly disagree"(or "rarely" and "never").
296 The highest and lowest agreements were for items "I will use it if it is reasonably priced" and "I
297 will buy when the first model is released", respectively.

298 Results of the structural equation modeling showed that all of the examined constructs had
299 significant effects on the AV acceptance. Most of the adaption models explained the variance in
300 acceptance of AVs less than 69% [62-72] although the proposed model explained 72.5% of the
301 variance in acceptance. Among the examined variables, only PR had an expected negative sign (-
302 0.161); PE and EC had the highest and lowest effects (0.215 & 0.044) on the AV acceptance,
303 respectively. Regression weights of DOI-related variables showed that OB had a greater effect
304 (0.094) than TR (0.067). Among variables related to the UTAUT theory, PE and SI (0.077) had
305 the highest and lowest effects on the AV, respectively. Among considered variables, except those
306 related to behavioral theories, PR had the highest and EC had the least impact on the acceptance
307 of AVs.

308 **5-1- Strategies and policy implications**

309 This study provides policymakers with several recommendations for allocation of resources in
310 promoting consumer acceptance of AVs. According to findings, it is suggested that necessary
311 conditions should be provided and following measures be taken for as many potential AV
312 customers to become actual users:

313 According to the significance of TR construct, it is recommended that designers and decision-
314 makers allow individuals to test AVs before purchasing/ intending to use them. In relation to OB,
315 it is suggested that stakeholders publish the performance reports of self-driving vehicles through
316 the social media and make a side by side comparison between self-driving cars and conventional
317 ones. In relation to EE, setting policies such as enabling their user-friendly designs so that people
318 feel comfortable when using their various features and/ or reduce the number of AVs'
319 components that need user-vehicle interaction are recommended. In case of PE, the policy of
320 enabling the comparison of AVs' significant advantages (less/optimal travel time, reduced fuel
321 consumption/costs, cost-effectiveness, comfort, etc.) over conventional cars could be considered.
322 Regarding the significance of SI, sharing the positive experiences of individuals (particularly the
323 celebrities) to their friends, colleagues and the social media could affect the acceptance of AVs.
324 To remove the safety concerns as a barrier of AVs' acceptance, designers should consider
325 subjects to promote vehicle's safety using preventing accidents and protecting their
326 lives/property. Besides, it is suggested that AVs run on exclusive lanes to improve the safety
327 perception of individuals. Reducing fuel consumption/ emissions through using compatible
328 alternative fuels is suggested to increase the acceptance of AVs by individuals with higher
329 environmental concerns. In relation to CI, Using attractive and up-to-date technologies can have
330 a significant influence in acceptance of AVs.

331

332 **5-2- Limitations and recommendations for further study**

333 Despite some policy implications for decision makers and designers, the findings should be
334 interpreted carefully. First: In this study, data was collected by questionnaires (as a conventional
335 method) at one point in time, which is a limitation because the method is not free from the
336 respondents' subjectivity. Future studies can use other objective qualitative data collection
337 methods such as actual experience of AVs to better understand the factors affecting the AV
338 acceptance.

339 Second: we used a stated preference questionnaire, since there was no implication of AV in Iran.
340 The results can be affected by hypothetical biases (individuals may report unrealistic values to
341 researchers). Further studies can compare results from the stated preference questionnaire to real
342 world results when self-driving cars have been implemented in Iran.

343 Third: Due to using of data collected in Tehran, implies that the results are not necessarily
344 applicable to other countries/cultures because of varying attitudes. However, the study raises
345 discussion points useful for future comparative studies aimed at exploring differences among
346 countries.

347 Many researchers believe that trust not only shapes inter-human relationships, but also affects
348 human-computer system relations [62]. They said that trust has three dimensions; one refers to a
349 person's belief that the system is able to understand and predict, the other states that technology
350 performs its assigned tasks accurately and correctly and the third refers to the belief that the
351 system provides enough and effective assistance to the individual [72]. It is suggested that future
352 studies examine the effects of the mentioned dimensions on the AV acceptance and on the PR.

353

354 **References**

355 1. Ghaffari Targhi, M. "Factors Influencing the Use of Autonomous and Shared
356 Autonomous Vehicles in Alberta" (Master's thesis, Graduate Studies) (2017).

357 2. Cheein, F. A. A., De La Cruz, C., Bastos, T. F., et al. "Slam-based cross-a-door solution
358 approach for a robotic wheelchair". *International Journal of Advanced Robotic Systems*,
359 **Vol. 6**, pp. 239-248 (2009). DOI: 10.5772/7230

360 3. Engelberg, J. K., Hill, L. L., Rybar, J., et al. "Distracted driving behaviors related to cell
361 phone use among middle-aged adults". *Journal of Transport & Health*, **Vol. 2**, pp. 434-
362 440 (2015) . DOI:10.1016/j.jth.2015.05.002

363 4. Fagnant, D. J., & Kockelman, K. "Preparing a nation for autonomous vehicles:
364 opportunities, barriers and policy recommendations". *Transportation Research Part A:
365 Policy and Practice*, **Vol. 77**, pp. 167-181 (2015) . DOI: 10.1016/j.tra.2015.04.003

366 5. Piao, J., McDonald, M., Hounsell, N., et al. "Public views towards implementation of
367 automated vehicles in urban areas". *Transportation research procedia*, **Vol. 14**, pp. 2168-
368 2177 (2016) . DOI: 10.1016/j.trpro.2016.05.232

369 6. Anderson, J. M., Nidhi, K., Stanley, K. D., et al. "Autonomous vehicle technology: A
370 guide for policymakers". Rand Corporation (2014).

371 7. Labi, S., Saeed, T. U., Volovski, M., et al. "An exploratory discussion of the impacts of
372 driverless vehicle operations on the man-made environment". In *1st International
373 Conference on Mechanical and Transportation Engineering*. Kuala Lumpur, Malaysia.
374 (2015).

375 8. Meyer, J., Becker, H., Bösch, P. M., et al. "Autonomous vehicles: The next jump in
376 accessibilities?". *Research in transportation economics*, **Vol. 62**, pp. 80-91 (2017). DOI:
377 10.1016/j.retrec.2017.03.005

378 9. Acheampong, R. A., & Cugurullo, F. "Capturing the behavioural determinants behind the
379 adoption of autonomous vehicles: Conceptual frameworks and measurement models to
380 predict public transport, sharing and ownership trends of self-driving cars".
381 *Transportation research part F: traffic psychology and behaviour*, **Vol. 62**, pp. 349-375
382 (2019) . DOI: 10.1016/j.trf.2019.01.009

383 10. Liu, F., Zhao, F., Liu, Z., et al. "Can autonomous vehicle reduce greenhouse gas
384 emissions? A country-level evaluation". *Energy Policy*, **Vol. 132**, pp. 462-473 (2019) .
385 DOI: 10.1016/j.enpol.2019.06.013

386 11. Severino, A., Curto, S., Barberi, S., Arena, F., et al. "Autonomous vehicles: An analysis
387 both on their distinctiveness and the potential impact on urban transport systems". *Appl.
388 Sci.* 2021, 11, 3604. DOI: 10.3390/app11083604

389 12. Lee, J., Lee, D., Park, Y., et al. "Autonomous vehicles can be shared, but a feeling of
390 ownership is important: Examination of the influential factors for intention to use
391 autonomous vehicles". *Transp. Res. Part C Emerg. Technol.* 2019, 107, 411–422. DOI:
392 10.1016/j.trc.2019.08.020

393 13. Taherdoost, H. "Importance of technology acceptance assessment for successful
394 implementation and development of new technologies". *Global Journal of Engineering
395 Sciences*, **Vol. 1**, pp. 1-3 (2019). DOI: 10.33552/GJES.2019.01.000511

396 14. Harst, L., Lantsch, H., & Scheibe, M. "Theories predicting end-user acceptance of
397 telemedicine use: systematic review". *Journal of medical Internet research*, **Vol. 21**, pp.
398 117- 131 (2019). DOI: 10.2196/13117

- 399 15. Thierer, A. "Technopanics, threat inflation, and the danger of an information technology
400 precautionary principle". *Minn. JL Sci. & Tech* (2013).
- 401 16. Thierer A & Hagemann R. "Removing roadblocks to intelligent vehicles and driverless
402 cars". mercatus working paper— Mercatus center George Mason University Washington
403 Vld, Virginia (2014).
- 404 17. Smith, B. W. "SAE levels of driving automation. Center for Internet and Society.
405 Stanford Law School". [http://cyberlaw.stanford.edu/blog/2013/12/sae-levels-](http://cyberlaw.stanford.edu/blog/2013/12/sae-levels-drivingautomation)
406 [drivingautomation](http://cyberlaw.stanford.edu/blog/2013/12/sae-levels-drivingautomation) (2013). DOI: 10.1016/j.jsr.2019.11.005
- 407 18. Farzin, I., Mamdoohi, A. R., & Ciari, F. "Autonomous Vehicles Acceptance: A Perceived
408 Risk Extension of Unified Theory of Acceptance and Use of Technology and Diffusion
409 of Innovation, Evidence from Tehran, Iran." *International Journal of Human-Computer*
410 *Interaction*, pp. 1-10. (2022). DOI: 10.1080/10447318.2022.2083464
- 411 19. Farzin, I., Abbasi, M., Macioszek, E., et al."Moving toward a More Sustainable
412 Autonomous Mobility, Case of Heterogeneity in Preferences". *Sustainability*, Vol. 15(1),
413 (2023). DOI: 10.3390/su15010460
- 414 20. Janatabadi, F., & Ermagun, A. "Empirical evidence of bias in public acceptance of
415 autonomous vehicles." *Transportation research part F: traffic psychology and behavior*,
416 Vol. 84, pp. 330-347 (2022). DOI: 10.1016/j.trf.2021.12.005
- 417 21. Ribeiro, M. A., Gursoy, D., & Chi, O. H. "Customer acceptance of autonomous vehicles
418 in travel and tourism." *Journal of Travel Research*. Vol. 61.3. pp. 620-636, (2022). DOI:
419 10.1177/0047287521993578
- 420 22. Miller, K., Chng, S., & Cheah, L. "Understanding acceptance of shared autonomous
421 vehicles among people with different mobility and communication needs." *Travel*
422 *Behaviour and Society*, Vol. 29, pp. 200-210 (2022) . DOI: 10.1016/j.tbs.2022.06.007
- 423 23. Mara, M., & Kathrin M. "Acceptance of autonomous vehicles: An overview of user-
424 specific, car-specific and contextual determinants." *User experience design in the era of*
425 *automated driving*. Pp. 51-83, (2022). DOI: 10.1007/978-3-030-77726-5_3
- 426 24. Gkartzonikas, C., Losada-Rojas, L. L., Christ, S., et al."A multi-group analysis of the
427 behavioral intention to ride in autonomous vehicles: evidence from three US metropolitan
428 areas." *Transportation*, pp. 1-41, (2022). DOI: 10.1007/s11116-021-10256-7
- 429 25. McLeay, F., Olya, H., Liu, H., et al."A multi-analytical approach to studying customers
430 motivations to use innovative totally autonomous vehicles." *Technological Forecasting*
431 *and Social Change*, Vol. 174 (2022). DOI: 10.1016/j.techfore.2021.121252
- 432 26. Chaveesuk, S., Chaiyasoonthorn, W., Kamales, N., et al."Evaluating the Determinants of
433 Consumer Adoption of Autonomous Vehicles in Thailand—An Extended UTAUT
434 Model. *Energies*, Vol. 16(2), (2023). DOI: 10.3390/en16020855
- 435 27. Jing, P., Xu, G., Chen, Y., et al."The determinants behind the acceptance of autonomous
436 vehicles: a systematic review". *Sustainability*, Vol. 12, pp. 1-26 (2020). DOI:
437 10.3390/su12051719
- 438 28. Venkatesh, V., Thong, J. Y., & Xu, X. "Consumer Acceptance and Use of Information
439 Technology: Extending the Unified Theory of Acceptance and Use of Technology". *MIS*
440 *Quarterly*, Vol. 36, pp. 157-178 (2012). DOI:10.2307/41410412
- 441 29. Taherdoost, H. "A review of technology acceptance and adoption models and theories.
442 *Procedia manufacturing*", Vol. 22, pp. 960-967 (2018). DOI:
443 10.1016/j.promfg.2018.03.137
- 444 30. Rogers, E. M. "Diffusion of innovations. Simon and Schuster" (2010)

- 445 31. Edrisi, A., & Ganjipour, H. "Exploring the acceptance of delivery robots by online buyers
446 using diffusion of innovation theory and structural equation modeling." *Amirkabir*
447 *Journal of Civil Engineering*. Vol. 54. pp. 14-24, (2022). DOI:
448 10.22060/CEEJ.2022.19258.7116
- 449 32. Tao, T., & Jason C. "Examining motivations for owning autonomous vehicles:
450 Implications for land use and transportation." *Journal of Transport Geography*, Vol.102,
451 (2022). DOI: 10.1016/j.jtrangeo.2022.103361
- 452 33. Xia, Z., Wu, D., & Zhang, L. "Economic, functional, and social factors influencing
453 electric vehicles' adoption: An empirical study based on the diffusion of innovation
454 theory." *Sustainability*, Vol. 14.10, (2022). DOI: 10.3390/su14106283
- 455 34. Yuen, K. F., Wong, Y. D., Ma, F., et al."The determinants of public acceptance of
456 autonomous vehicles: An innovation diffusion perspective". *Journal of Cleaner*
457 *Production*, **Vol. 270**, pp. 121904 (2020). DOI: 10.1016/j.jclepro.2020.121904
- 458 35. Nordhoff, S., Malmsten, V., van Arem, B., et al."A structural equation modeling
459 approach for the acceptance of driverless automated shuttles based on constructs from the
460 Unified Theory of Acceptance and Use of Technology and the Diffusion of Innovation
461 Theory". *Transportation Research Part F: Traffic Psychology and Behaviour*, **Vol. 78**, pp.
462 58-73 (2021). DOI: 10.1016/j.trf.2021.01.001
- 463 36. Venkatesh, V., Morris, M. G., Davis, G. B., et al."User acceptance of information
464 technology: Toward a unified view". *MIS quarterly*, 425-478 (2003). DOI:
465 10.2307/30036540
- 466 37. Ingeveld, M. "Usage intention of automated vehicles amongst elderly in the Netherlands"
467 (Master's thesis, Graduate Studies) (2017).
- 468 38. Leicht, T., Chtourou, A., & Youssef, K. B. "Consumer innovativeness and intentioned
469 autonomous car adoption". *The Journal of High Technology Management Research*, **Vol.**
470 **29**, pp. 1-11 (2018). DOI: 10.1016/j.hitech.2018.04.001
- 471 39. Madigan, R., Louw, T., Wilbrink, M., et al. "What influences the decision to use
472 automated public transport? Using UTAUT to understand public acceptance of automated
473 road transport systems". *Transportation research part F: traffic psychology and*
474 *behaviour*, **Vol. 50**, pp. 55-64 (2017). DOI: 10.1016/j.trf.2017.07.007
- 475 40. Schierz, P. G., Schilke, O., & Wirtz, B. W. "Understanding consumer acceptance of
476 mobile payment services: An empirical analysis". *Electronic commerce research and*
477 *applications*, **Vol. 9**, pp. 209-216 (2010). DOI: 10.1016/j.elerap.2009.07.005
- 478 41. Martins, C., Oliveira, T., & Popovič, A. "Understanding the Internet banking adoption: A
479 unified theory of acceptance and use of technology and perceived risk application".
480 *International journal of information management*, **Vol. 34**, pp. 1-13 (2014). DOI:
481 10.1016/j.ijinfomgt.2013.06.002
- 482 42. Ross, I. "Perceived risk and consumer behavior: a critical review". *ACR North American*
483 *Advances* (1975).
- 484 43. Guériau, M., Cugurullo, F., Acheampong, R. A., et al. "Shared autonomous mobility on
485 demand: A learning-based approach and its performance in the presence of traffic
486 congestion". *IEEE Intell. Transp. Syst. Mag.* 2020, 12, 208–218. DOI:
487 10.1109/MITS.2020.3014417
- 488 44. Campisi, T., Severino, A., Al-Rashid, M. A., et al. "The development of the smart cities
489 in the connected and autonomous vehicles (CAVs) era: From mobility patterns to scaling
490 in cities". *Infrastructures* 2021, 6, 100. DOI: 10.3390/infrastructures6070100

- 491 45. Simko, D.J. Increasing Road Infrastructure Capacity through the Use of Autonomous
492 Vehicles; Naval Postgraduate School: Monterey, CA, USA, 2016
- 493 46. Featherman, M. S., & Pavlou, P. A. "Predicting e-services adoption: a perceived risk
494 facets perspective". *International journal of human-computer studies*, **Vol. 59**, pp. 451-
495 474 (2003). DOI: 10.1016/S1071-5819(03)00111-3
- 496 47. Ko, H., Jung, J., Kim, J., et al."Cross-cultural differences in perceived risk of online
497 shopping". *Journal of Interactive Advertising*, **Vol. 4**, pp. 20-29 (2004). DOI:
498 10.1080/15252019.2004.10722084
- 499 48. Zmud, J., Sener, I. N., & Wagner, J. "Consumer acceptance and travel behavior: impacts
500 of automated vehicles". Texas A&M Transportation Institute (2016).
- 501 49. Liu, H., Yang, R., Wang, L., et al."Evaluating initial public acceptance of highly and
502 fully autonomous vehicles". *International Journal of Human-Computer Interaction*, **Vol.**
503 **35**, pp. 919-931 (2019). DOI: 10.1080/10447318.2018.1561791
- 504 50. Yeung, S. P. M. "Teaching approaches in geography and students' environmental
505 attitudes". *Environmentalist*, **Vol. 24**, pp. 101-117 (2004). DOI: 10.1007/s10669-004-
506 4801-1
- 507 51. Rehman, Z. U., & Dost, M. K. "Conceptualizing green purchase intention in emerging
508 markets: An empirical analysis on Pakistan". In *The 2013 WEI International Academic*
509 *Conference Proceedings*, **Vol. 1**, pp. 101-112 (2013).
- 510 52. do Paço, A. & Raposo, M. "'Green' segmentation: an application to the Portuguese
511 consumer market", *Marketing Intelligence & Planning*, **Vol. 27**, pp. 364-37 (2009). DOI:
512 10.1108/02634500910955245
- 513 53. Ross, C., & Guhathakurta, S.. "Autonomous vehicles and energy impacts: a scenario
514 analysis". *Energy Procedia*, **Vol. 143**, pp. 47-52 (2017). DOI:
515 10.1016/j.egypro.2017.12.646
- 516 54. Sreejesh, S. "Consumers' Evaluation of Brand Extensions: An Application of Multiple-
517 Group Causal Models in Assessing Cross Product Category Measurement Equivalence".
518 *Southern Business Review*, **Vol. 36**, pp. 1-23 (2011).
- 519 55. Hirschman, E. C. "Innovativeness, novelty seeking, and consumer creativity". *Journal of*
520 *consumer research*, **Vol. 7**, pp. 283-295 (1980). DOI: 10.1086/208816
- 521 56. Xie, Y. H. "Consumer innovativeness and consumer acceptance of brand extensions".
522 *Journal of Product & Brand Management* (2008). DOI: 10.1108/10610420810887581
- 523 57. Haboucha, C. J., Ishaq, R., & Shiftan, Y."User preferences regarding autonomous
524 vehicles". *Transportation Research Part C: Emerging Technologies*, **Vol. 78**, pp. 37-49
525 (2017). DOI: 10.1016/j.trc.2017.01.010
- 526 58. Mueller, R. O. "Basic principles of structural equation modeling: An introduction to
527 LISREL and EQS". Springer Science & Business Media (1999). DOI:
528 10.1080/10548408.2018.1507866
- 529 59. Hair, J. F. "Multivariate data analysis". Upper Saddle River, NJ [etc.]. Pearson Prentice
530 Hall, New York, NY: Macmillan (2009).
- 531 60. Al- Gahtani, S. S. "Computer technology adoption in Saudi Arabia: Correlates of
532 perceived innovation attributes". *Information Technology for Development*, **Vol.10**,
533 pp.57-69 (2003). DOI: 10.1002/itdj.1590100106
- 534 61. Mazzarol, T., & Reboud, S. "Adoption and Diffusion of Innovation. In *Entrepreneurship*
535 *and Innovation*". Springer, Singapore (2020). DOI: 10.1007/978-981-13-9412-6_6

- 536 62. Choi, J. K., & Ji, Y. G. "Investigating the importance of trust on adopting an autonomous
537 vehicle". *International Journal of Human-Computer Interaction*, **Vol. 31**, pp. 692-702
538 (2015). DOI: 10.1080/10447318.2015.1070549
- 539 63. Liu, P., Yang, R., & Xu, Z. "Public acceptance of fully automated driving: Effects of
540 social trust and risk/benefit perceptions". *Risk Analysis*, **Vol. 39**, pp. 326-341 (2019).
541 DOI: 10.1111/risa.13143
- 542 64. Mohtar, S., & Abbas, M. "Consumer resistance to innovation due to perceived risk:
543 Relationship between perceived risk and consumer resistances to innovation". *Journal of
544 Technology and Operations Management*, **Vol. 10**, pp. 1-13 (2020). DOI:
545 10.32890/jtom2015.10.1.1
- 546 65. Fransson, N., & Gärling, T. "Environmental concern: Conceptual definitions,
547 measurement methods, and research findings". *Journal of environmental psychology*,
548 **Vol. 19**, pp. 369-382 (1999). DOI: 10.1006/jevp.1999.0141
- 549 66. Gkartzonikas, C., & Gkritza, K. "What have we learned? A review of stated preference
550 and choice studies on autonomous vehicles". *Transportation Research Part C: Emerging
551 Technologies*, **Vol. 98**, pp. 323-337 (2019). DOI: 10.1016/j.trc.2018.12.003
- 552 67. Roehrich, G. "Consumer innovativeness: Concepts and measurements". *Journal of
553 business research*, **Vol. 57**, pp. 671-677 (2004). DOI: 10.1016/S0148-2963(02)00311-9
- 554 68. Moore, G. C., & Benbasat, I. "Development of an instrument to measure the perceptions
555 of adopting an information technology innovation". *Information systems research*, **Vol. 2**,
556 pp. 192-222 (1991). DOI: 10.1287/isre.2.3.192
- 557 69. Tran, T. C. T., & Cheng, M. S. "Adding Innovation Diffusion Theory to Technology
558 Acceptance Model: Understanding Consumers' Intention to Use Biofuels in Viet Nam".
559 *International review of management and business research*, **Vol. 6(2)**, pp. 595-615
560 (2017).
- 561 70. Yuen, K. F., Wang, X., Ng, L., et al. "An investigation of customers' intention to use self-
562 collection services for last-mile delivery". *Transport Policy*, **Vol. 66**, pp. 1-8 (2018).
563 DOI: 10.1016/j.tranpol.2018.03.001
- 564 71. Osswald, S., Wurhofer, D., Trösterer, S., et al. "Predicting information technology usage
565 in the car: towards a car technology acceptance model". In *Proceedings of the 4th
566 International Conference on Automotive User Interfaces and Interactive Vehicular
567 Applications* (pp. 51-58) (2012). DOI: 10.1145/2390256.2390264
- 568 72. Min, S., So, K. K. F., & Jeong, M. "Consumer adoption of the Uber mobile application:
569 Insights from diffusion of innovation theory and technology acceptance model". *Journal
570 of Travel & Tourism Marketing*, **Vol. 36**, pp. 770-783 (2019).
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Figure and table captions

Fig. (1): Pie chart to show the research variables (inner ring), classification based on internal and external (middle ring) and derived from behavioral or non-behavioral theories (outer ring)

Fig. (2): Proposed conceptual model (Based on the literature review, H+&H- are the hypotheses related to the positive & negative effects of the independent variable on the dependent variable, respectively)

ACC: Acceptance of autonomous vehicle; TR: Trialability; OB: Observability; PE: Performance expectancy; EE: Effort expectancy; SI: Social influence; PR: Perceived risk; CI: Consumer Innovativeness; EC: Environmental concern

Figure (3) Data flow chart for current research

Fig. (4): Statistical analyses results of items related to the acceptance of autonomous vehicles

Fig. (5): Regression weights and evaluation criteria of the structural model (ACC: Acceptance of autonomous vehicle; TR: Trialability; OB: Observability; PE: Performance expectancy; EE: Effort expectancy; SI: Social influence; PR: Perceived risk; CI: Consumer Innovativeness; EC: Environmental concern)

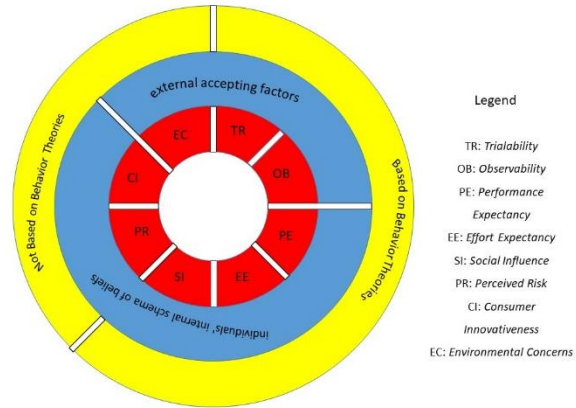
Table (1): Participants' profile of survey

Table (2): Standard regression weights and evaluation criteria of measurement model

Table (3): Survey validation and model fit

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Figures and tables



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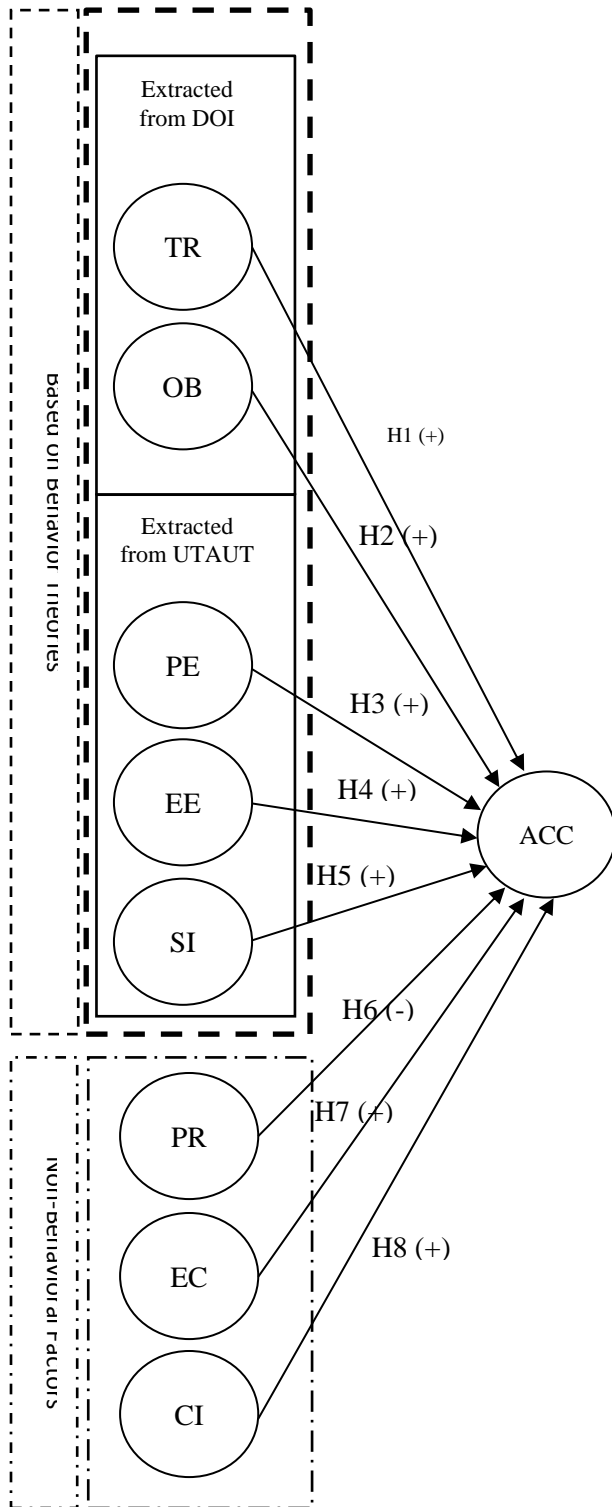
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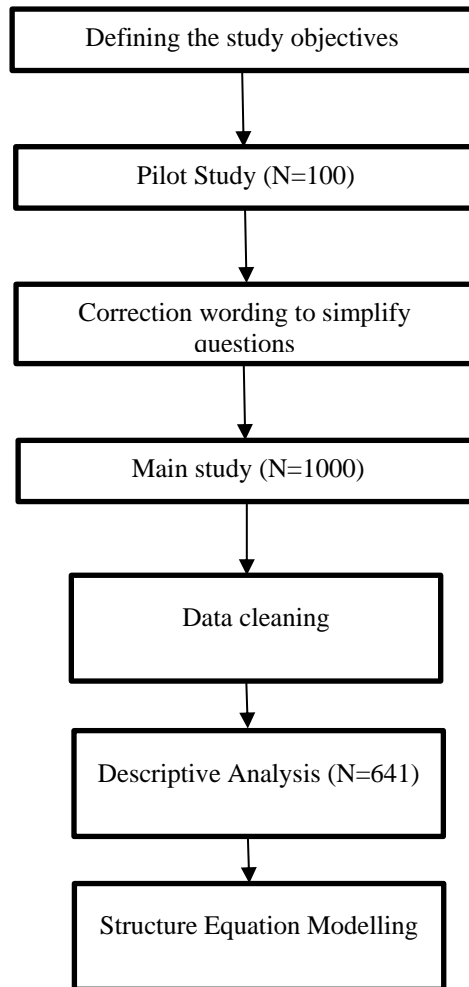


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 599 **related to the positive & negative effects of the independent variable on the dependent variable,**
 600 **respectively)**

601 **ACC: Acceptance of autonomous vehicle; TR: Trialability; OB: Observability; PE: Performance**
 602 **expectancy; EE: Effort expectancy; SI: Social influence; PR: Perceived risk; CI: Consumer**
 603 **Innovativeness; EC: Environmental concern**

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Figure (3) Data flow chart for current research

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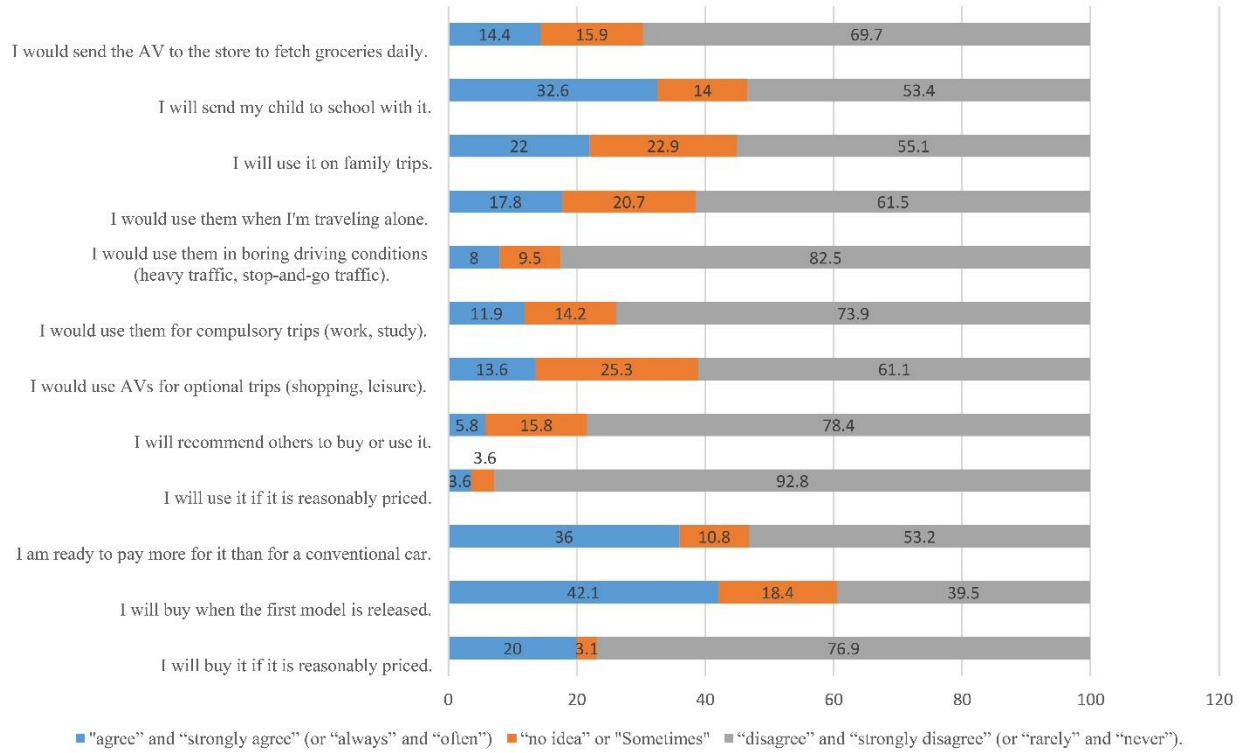
Table (1): Participants' profile of survey

Variable	Category	Frequency	Percent	Average	Standard Deviation
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Gender	Male	292	45.6	0.54	0.498
	Female	349	54.4		
Marital Status	Single	281	43.8	0.56	0.497
	Married	360	56.2		
Age	<14	1	0.2	3.93	0.730
	15-24	181	28.2		
	25-44	330	51.4		
	45-64	119	18.6		
	+65	10	1.6		
Education	Lower than Diploma	65	10.1	2.70	0.808
	Diploma	141	22.0		
	Associate Degree and Bachelor of Science	359	56.0		
	Master of Science and Doctorate	76	11.9		
Household Size	1	14	2.2	3.82	1.241
	2	66	10.3		
	3	160	25.0		
	4	257	40.1		
	5+	144	22.4		

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Fig. (4): Statistical analyses results of items related to the acceptance of autonomous vehicles

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Table (2): Standard regression weights and evaluation criteria of measurement model

Latent Variable	Item	Standard Regression Weight	Source
Trialability	I prefer to try them long enough.	0.77	[62]
	The ability to try them would be useful in my decision to use them.	0.91	[63]
	The ability to try them would be useful in my decision to buy them.	0.88	
Observability	In the future, by watching others use them, I will be able to.....		[64]
	a) Learn how they work.	0.82	
	b) Explain how they work to others.	0.78	
	c) I can say if they are useful to me.	0.89	
Performance Expectancy	d) Clearly understand how they work.	0.89	[65]
	Due to the use of technology and effective communication with other vehicles, I can reach my	0.74	

	destination faster.		
	AVs would enhance my performance while driving because I would be able to do other things (eating, sleeping, and using a computer).	0.67	
	They are easier to use and better than conventional cars.	0.79	[66]
	Overall, they are a good transportation alternative.	0.78	
	I would be able to easily adjust my daily schedule using my AV.	0.76	[67]
Effort Expectancy	It would be easy for me to use them to accomplish my goals.	0.83	[66]
	It would be easy for me to learn how to use AVs.	0.81	[67]
	I will not need much mental effort to interact with it.	0.71	[62]
Social Influence	Individuals who are important to me will think I should use them too.	0.80	[37]
	People will successfully accept it because its use looks good to others.	0.56	
	The people whose opinions I care about would encourage me to use AVs.	0.86	[39]
	People who influence my behavior will encourage me to use them.	0.95	
Perceived Risk	I am generally worried about using them.	0.85	[68]
	I am concerned about their safety.	0.78	
	I am concerned about the shared use of transport infrastructure by autonomous and conventional vehicles.	0.63	[10]
Environmental Concerns	We need more and better public transportation even if it means more taxes.	0.53	[69]
	We must decide and act on controlling greenhouse gas emissions.	0.56	
	I would like to pay more to buy products that are more environmentally friendly.	0.90	[57]

	I experience new technology products earlier than people around me.	0.91	
Consumer Innovative	I am aware of the latest technologies more than others.	0.76	[57]
	I often buy new technologies even if they are expensive.	0.65	
	Most technologies are great.	0.61	
	I will use it if it is reasonably priced.	0.60	[70]
Acceptance of AV	I would use AVs for optional trips (shopping, leisure).	0.56	
	I would use them in boring driving conditions (heavy traffic, stop-and-go traffic).	0.67	Created for the present study
	I will send my child to school with it.	0.56	
	I would send the AV to the store to fetch groceries daily.	0.58	

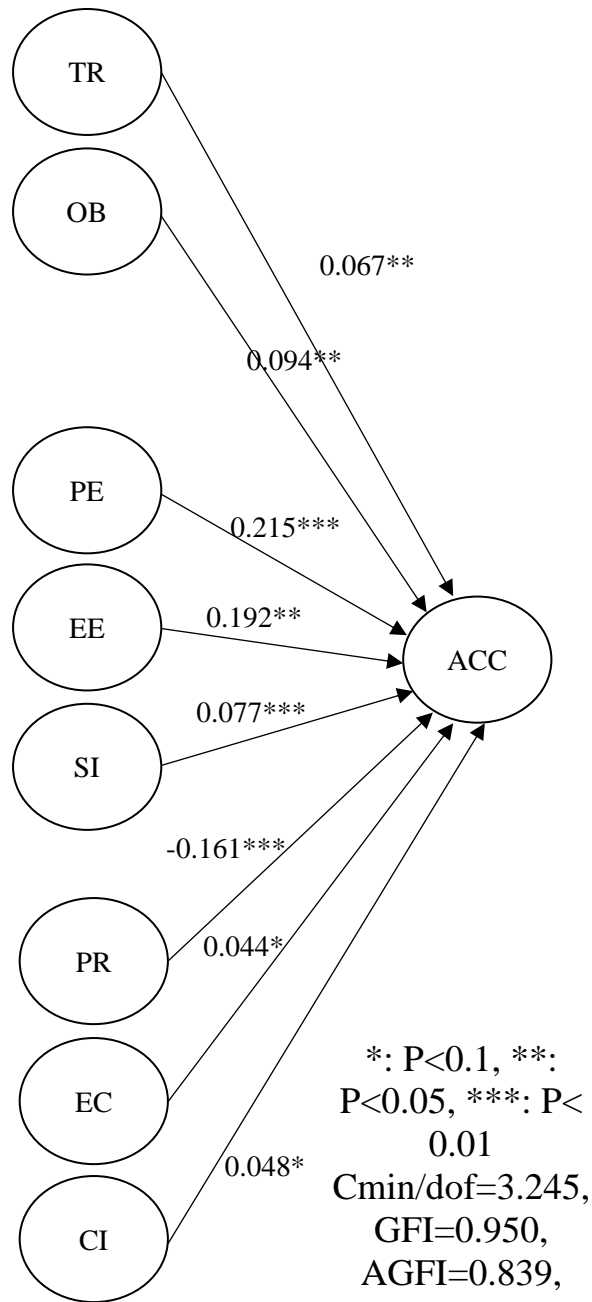
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621 **Fig. (5): Regression weights and evaluation criteria of the structural model (ACC: Acceptance of**
 622 **autonomous vehicle; TR: Trialability; OB: Observability; PE: Performance expectancy; EE: Effort**
 623 **expectancy; SI: Social influence; PR: Perceived risk; CI: Consumer Innovativeness; EC:**
 624 **Environmental concern)**

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Table (3): Survey validation and model fit

Latent Variable	Cronbach's Alpha	Construct Reliability (CR)	Average Variance) Extract (AVE)
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Trialability	0.872	0.729	0.889
Observability	0.906	0.715	0.909
Performance Expectancy	0.846	0.560	0.864
Effort Expectancy	0.826	0.615	0.827
Social Influence	0.843	0.651	0.878
Perceived Risk	0.848	0.575	0.800
Environmental Concerns	0.778	0.526	0.762
Consumer Innovative	0.799	0.544	0.823
Acceptance of AV	0.722	0.505	0.835

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Authors:

Iman Farzin:

Iman Farzin graduated from Isfahan University of Technology in civil engineering, and earned an MSc degree in Transportation Planning from Tarbiat Modares University. He received his PhD in transportation planning from Tarbiat Modares University, Tehran, Iran in 2021. His research interests are discrete choice modeling, behavioral models in transportation, supply chain, and travel demand management.

Amir Reza Mamdoohi:

Amir Reza Mamdoohi is associate professor at the faculty of civil and environmental engineering, Tarbiat Modares University, Tehran, Iran. He has supervised and co-advised more than 10 PhD and 100 MS theses, published more than 115 journal papers and more than 120 conference papers. His international collaboration has resulted in signing memorandum of understanding (MOUs) with TU Wien (Austria), TU Berlin (Germany) and IST (Portugal) and action plans including joint international workshops, seminars, theses co-advisorship, projects and papers. His international projects include Innovative Traffic Congestion Alleviation for Dubai, a Proposed Short-Term Action Plan, Blue City Master Plan Revision and Sustainable Urban Mobility Research in Central Asia (SUMRICA).

Francesco Ciari:

Francesco Ciari obtained his master degree in Environmental engineering at the University of Florence in 2003. He obtained his PhD in transportation planning in 2012 with a dissertation titled Sharing as a key to rethink urban mobility at the Swiss Federal Institute of Technology (ETH) in Zurich, where he worked as a senior researcher until 2017. Between 2017 and 2018, he joined Joanneum Research in Graz (Austria) as head of the Urban Living Lab research unit. Currently he is associate professor in department of civil, geological and mining engineering, Polytechnique Montréal University, Montréal, Canada