

1 **Can a unique appearance of e-bikes, coupled with information on their** 2 **characteristics, influence drivers' gap acceptance?**

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19 **ABSTRACT**

20 *Objective:* Earlier studies have shown that, when confronted with approaching e-bikes, car drivers tend to
21 underestimate their speed and accept smaller gaps for crossing in front of them compared to conventional
22 bicycles (Petzoldt et al. 2017b; Schleinitz et al. 2016b). As an explanation, it has been suggested that car
23 drivers rely on their previous experience with conventional bicycles, which tells them that those mostly
24 travel at low speeds. E-bikes, which look just like regular bicycles, do not conform to this expectation,
25 resulting in potentially dangerous interactions. Based on this assumption, researchers have suggested to
26 increase other road users' awareness of e-bikes' higher speeds by giving them a distinct appearance. The
27 goal of our experiment was to investigate the effects of such a unique appearance, aided by clear instructions
28 about the higher speeds of e-bikes, on gap acceptance.

29 *Method:* In order to investigate the effect of appearance independent of the effect of bicycle type, we used
30 video sequences of conventional bicycles and e-bikes approaching at different levels of speed. The riders
31 (regardless of what type of bike they were actually riding) either wore an orange helmet as an indicator for
32 an e-bike, or a grey helmet indicating a conventional bicycle. Fifty participants were asked to indicate the
33 smallest acceptable gap for a left turn in front of the cyclist or e-bike rider.

34 *Results:* The results showed significantly smaller acceptable gaps when confronted with the grey helmet
35 (signal for bicycle) compared to the orange helmet (signal for e-bike), whereas there was no difference
36 between the actual bicycle types.

37 *Conclusions:* Overall, the results indicate that informing about e-bikes characteristics in combination with
38 a unique appearance can lead to a more cautious behaviour among car drivers.

39
40 **Keywords:** time to collision, gap acceptance, electric bicycle, speed judgements.

41 INTRODUCTION

42 As the number of e-bikes is growing considerably on roads all over the world (Confederation of European
43 Bicycle Industry 2017; Rose 2012), new problems for traffic safety arise. Various studies have shown that
44 e-bike riders travel at higher speeds than conventional cyclists (Huertas-Leyva et al. 2018; Schleinitz et al.
45 2017; Vlakveld et al. 2015). It has been suggested that this could cause safety critical situations, as other
46 road users might misperceive an e-bike for a conventional bicycle, and as a consequence, underestimate its
47 speed (Jellinek et al. 2013; Popovich et al. 2014; Scaramuzza et al. 2015). This is corroborated by e-bike
48 riders themselves, one of whom explained that “you ride faster on an e-bike and often car drivers are not
49 aware of that. Especially, when drivers intend to turn right they think they can manage before one comes
50 past on a bike. I experienced that twice, where I had to break [sic] heavily” (Haustein and Møller, 2016).

51 As an explanation for the misperception of e-bikes' speed, it has been suggested that other road users have
52 learned that bicycles mostly travel at low speeds (Haustein and Møller, 2016). Based on prolonged experi-
53 ence, road user build heuristics – rules of thumb - about characteristics of vehicles. For example, it has been
54 shown that observers tend to judge specific car types as being faster than others (Davies 2009, Davies and
55 Patel 2005). If a new vehicle type is introduced, no heuristics based on previous experience exist, which,
56 in theory, should result in a more careful behaviour of other road users around such a vehicle, and a more
57 thorough observation and information processing. In the case of the e-bike, however, there are heuristics
58 available for a vehicle that looks like an e-bike - a conventional bicycle - however has different behavioural
59 characteristics (low speed, low acceleration). The fact that e-bikes do not conform to the expectation of low
60 speed and acceleration can therefore result in potentially dangerous interactions with other road users and
61 a higher crash risk. Indeed, when confronted with an approaching e-bike, car drivers tend to underestimate
62 their speed and accept smaller gaps for turning in front of them in comparison to conventional bicycles
63 (Petzoldt et al. 2017b; Schleinitz et al. 2016b). The authors argued that drivers relied on their previous
64 experience with bicycles, which led them to assess speed (to some degree) based on easily observable
65 features such as pedalling frequency or posture (as an indicator of effort), a strategy that is misleading in
66 the case of the e-bike, which allows for rather effortless cycling with comparatively low pedalling frequen-
67 cies. Indeed, when manipulated experimentally, pedalling frequency, and not bicycle type, was found to
68 have an effect on estimated gap size (Schleinitz et al. 2016b). The issue is complicated further by the fact
69 that approach speed as such already has an impact on accepted gap size, with smaller gaps being accepted
70 with increasing speed (Alexander et al. 2002; Cooper et al. 1977). Likewise, it has been found that vehicle
71 size has an effect on drivers' gap acceptance, with smaller gaps being accepted for smaller vehicles (Alex-
72 ander et al. 2002; Keskinen, et al. 1998).

73 Of course, it can be assumed that, after a longer period of exposure to and experience with e-bikes, other
74 road users would adapt, i.e. they learn that the vehicle that looks like a bicycle could actually as well be an
75 e-bike, and expect a larger range of speeds and accelerations. Simply relying on this process, however,
76 might be considered somewhat unethical, given the rapid increase of the number of e-bikes on the roads,
77 and the associated increase of crash risk in the short term. Researchers have therefore suggested to increase
78 other road users' awareness of e-bikes' higher speeds by giving them a distinct appearance (Dozza et al.
79 2016; Huertas-Leyva et al. 2018) or inform them about the higher speeds of e-bikes through campaigns

80 (Jellinek et al. 2013; Scaramuzza et al. 2015). Both of these aspects are relevant. While the knowledge of
81 e-bike's potential to reach higher levels of speed is vital to form appropriate expectations, it is equally
82 important to be able to distinguish between e-bikes and conventional bicycles, to be able to act on these
83 expectations. While previous attempts have produced only mixed results (Schleinitz and Petzoldt 2017),
84 the general approach of making e-bikes easily recognisable by a unique appearance and informing people
85 about the higher speed of the e-bike as well as their appearance clearly appears to be reasonable.

86 The goal of our experiment was to investigate the effects of a unique appearance for e-bikes, aided by clear
87 instructions about their potential to reach higher speeds, on drivers' gap acceptance. In order to investigate
88 this independent of the effect of bicycle type, we used video sequences of conventional bicycles and e-
89 bikes approaching at different levels of speed. To create the unique appearance, riders either wore an orange
90 helmet as an indicator for an e-bike, or a grey helmet indicating a conventional bicycle (regardless of what
91 type of bike they were actually riding).

92 **METHOD**

93 **Participants**

94 In total 50 participants took part in the experiment (16 males, 34 females). They were on average 22.8 years
95 old ($SD = 3.8$). More than 80% of the participants were students of Chemnitz University of Technology,
96 whereas the other pursued various professions. All of them had a drivers' license, were active drivers and
97 drove on average 5,482 km per year ($SD = 6,815$). They had normal or corrected visual acuity. For their
98 participation, they received monetary compensation.

99 **Experimental design**

100 We conducted a video-based laboratory experiment with scenes of different bicycle types approaching an
101 observer. The videos showed two different bicycle types: a conventional trekking bicycle (Diamant Ubari
102 black) and a comparable e-bike (Diamant Supreme), which approached at constant speeds from 15 to 35
103 km/h (in 5 km/h increments). In half of the videos, the rider wore either an orange helmet, in the other half
104 a grey one (independent of bicycle type). This resulted in 20 within-factor level combinations (5 speed
105 levels x 2 bicycle types x 2 helmet colours), for which the size of the minimum acceptable gap as indicated
106 by the participants was measured.

107 **Material**

108 Video material was recorded on a straight taxiway of a small general aviation airport. The point of view
109 was that of a driver, i.e. the height of the camera position was comparable to the eye level of a driver sitting
110 in a car. The observer/camera was put at a T-junction, in a position that is similar to a car waiting to make
111 a left turn. A cyclist approached at a constant speed in the oncoming lane across which the left turn would
112 have to be executed. The rider approached either on a conventional bicycle or on an e-bike. The position of
113 a potential collision between the approaching cyclist and the observer's car when turning left was marked
114 by a white line that was pasted onto the road.

115 After recording, all videos were edited to provide the participants a clear visual differentiation between
116 conventional bicycle and e-bike. In the original video material, the rider wore a grey helmet in all ap-
117 proaches. Within these videos, an orange helmet was “created” with the help of the video editing software
118 Sensarea 1.10.1 (see Figure 1). This was done with all videos for all combinations of speed and bicycle
119 type. This resulted in two sets of videos that were completely identical in all aspects except the colour of
120 the rider’s helmet (which was intended as an indicator of whether the rider was approaching on an e-bike
121 or a conventional bicycle).

122 All videos were cut so that the approaching cyclist was always in a distance of 100 m from the white line
123 at the start of each clip. The video continued until the vehicle had passed the position of the observer, which
124 resulted in variable clip length, dependent on approach speed. The video material was presented on a 23
125 inch screen. Participants were seated at a desk in a distance of 60 cm.

126 [Figure 1]

127 Procedure

128 At the beginning of the experiment, participants were informed about the differences between e-bikes and
129 conventional bicycles. First, it was explained that e-bikes support the riders while pedalling with an electric
130 motor up to 25 km/h, so that the rider can reach higher speeds than conventional bicycles. Second, they
131 received explicit information that in the videos, the e-bike rider would wear an orange helmet to better
132 distinguish him from the conventional cyclist. After these general instructions, the specific instructions on
133 gap acceptance followed. Participants should put themselves in the position of a car driver at the intersec-
134 tion, waiting to make a left turn. Their task was to indicate the minimum gap that they felt was acceptable
135 to complete a left turn in front of the approaching cyclist. They should indicate the moment the gap between
136 observer and cyclist had reached this critical size by pressing the spacebar. All video clips were shown
137 twice in randomized order, which resulted in 40 trials. The participants completed two practice trials before
138 data collection started. Afterwards, they had to fill in a short questionnaire on demographic variables (e.g.
139 age, gender, distance travelled per year). The experiment lasted about 20 - 30 minutes.

140 Data analysis

141 For data analysis, a mean value for minimum accepted gap size was calculated for each factor level com-
142 bination (helmet colour x bicycle type x speed). The statistical analysis of minimum accepted gap size was
143 conducted using a three-factor analysis of variance (ANOVA) for repeated measurements. Bonferroni cor-
144 rection was used for all pairwise comparisons.

145 RESULTS

146 One participant had to be removed from the dataset as he produced implausible values for accepted gap
147 size (gaps of only a few ms). Figure 2 depicts the minimum accepted gap size dependent on helmet colour
148 and approach speed, for the actual bicycle and Figure 3 for the actual e-bike. Table 1 shows the main effects
149 and interactions between the factor combinations as uncovered by the ANOVA. In the appendix, histograms
150 detailing the distribution of responses for the different levels of approach speed and the different helmet
151 colours can be found (Appendix, Figure A1-A5).

152 As Figure 2 and 3 show, the rider's approach speed had a strong (and significant, see Table 1) impact on
153 minimum accepted gap size, with smaller gaps being accepted at higher approach speeds. Post hoc tests
154 revealed significant differences between all speed levels (all $p < .001$). It seems that, while, surely, time to
155 collision (the time remaining until the approaching bicycle would reach the observer) is the single most
156 safety relevant characteristic of the approaching bicycle, it hardly seemed to play a role in observers deci-
157 sions. There also was a significant main effect of helmet colour, with the accepted gaps for the left turn
158 when the rider wore a grey helmet (supposed conventional bicycle; $M = 7.2$, $SD = 2.6$) being smaller than
159 when the helmet was orange (supposed e-bike; $M = 7.4$, $SD = 2.5$). It appears that participants were willing
160 to accept smaller gaps in front of what they believed was a bicycle, as compared to the supposed e-bike. In
161 contrast, we found no significant effect of actual bicycle type. The accepted gap size in front of the e-bike
162 rider ($M = 7.3$, $SD = 2.6$) and the conventional cyclist ($M = 7.3$, $SD = 2.6$) were identical.

163 In addition, the ANOVA revealed a significant interaction between helmet colour and speed. As can be
164 seen in Figure 2, the difference between the two helmet colours was reduced, and even disappeared, for the
165 higher speed levels, especially at 35 km/h ($M_{orange\ 35\ km/h} = 5.2$, $SD_{orange\ 35\ km/h} = 1.7$; $M_{grey\ 35\ km/h} = 5.3$, $SD_{grey\ 35\ km/h} = 1.6$).

167 [Figure 2 and 3]

168 [Table 1]

169 For descriptive purposes, we transformed the minimum accepted gap size from a time-based into a distance-
170 based metric (i.e., "how far is the bike away" instead of "when does it arrive"; see Figure 4 and 5). This
171 transformation leaves the main effect of helmet colour untouched (just as the non-effect of bicycle type).
172 Important, however, is the fact that there still is an effect of speed on the distance-based size of the accepted
173 gaps. Although observers did not seem to use time to collision as a basis for their decisions (Figure 2 and
174 3), they also did not just rely on a certain physical distance. It appears that they, at least to some degree,
175 accounted for the bicycles' approach speed. It should be noted, however, that this relationship does not
176 appear to be fully linear, as especially at the higher speed levels (30 km/h and 35 km/h), the differentiation
177 disappeared.

178 [Figure 4 and 5]

179 DISCUSSION

180 Goal of our experiment was to investigate whether it would be possible to influence drivers' gap acceptance
181 in front of approaching cyclists by introducing different appearances for the cyclists (i.e., helmets of differ-
182 ent colours), and linking those appearances to certain characteristics of the bicycle (i.e., e-bike vs. conven-
183 tional bicycle, and their potential to reach certain levels of speed). Our results show that, independent of
184 which type of bicycle type was actually approaching, minimum accepted gaps were larger in front of the
185 rider that was assumed (based on his appearance) to be riding an e-bike compared to the supposed cyclist,
186 although it should be noted that the difference is less than a second in practice. Still, these results suggest
187 that, in principle, a unique appearance coupled with clear instructions can influence drivers' gap acceptance,
188 so that a more cautious behaviour would be provoked.

189 In addition, an interaction effect between appearance and approach speed was found, as the effect of the
190 helmet colour on accepted gap size seemed to disappear at higher speed levels. A potential explanation can
191 be derived from investigations of the judgment of time to collision (Cavallo and Laurent 1988; McLeod
192 and Ross 1983), who reported that the judgment improves with higher approach speed. For our experiment,
193 it might be deduced, that, because the assessment of approach speed was easier to accomplish at the higher
194 speed levels, the impact of previous experience on gap acceptance is diminished.

195 Interestingly, the effect of actual bicycle type that had been found previously (Petzoldt et al. 2017b; Schlei-
196 nitz et al., 2016b) disappeared. It seems that, when observers are provided with a clear and unambiguous
197 cue as to what type of bicycle is approaching them, the role of other, less reliable cues (e.g., pedalling
198 frequency) is reduced. The unique appearance coupled with information about e-bikes' higher speed appar-
199 ently overcame the effect of heuristics and prior knowledge on conventional cyclists' speed.

200 In line with previous studies, an increase in speed led to shorter accepted time gaps in front of the bicycle
201 and the e-bike (Petzoldt et al., 2017b; Schleinitz et al. 2016a), an effect that was much stronger than the
202 effect of the rider's appearance (with a difference of only 0.2s on average). As several studies have shown
203 that e-bikes are indeed used at higher speeds levels than conventional bicycles (Huertas-Leyva et al. 2018;
204 Schleinitz et al. 2017; Vlakveld et al. 2015), a unique appearance might therefore not be fully sufficient to
205 counter a potential increase in crash risk for e-bikes. Compared to conventional cyclists, e-bike riders were
206 found to be especially at higher risk of safety critical situations at intersections (Dozza et al. 2016), e.g.,
207 when car drivers did not yield right-of-way to an e-bike (Petzoldt et al. 2017a; Schleinitz et al.
208 2014), presumably because they underestimated the e-bike's speed. A unique appearance alone might not
209 be enough to fully eradicate this type of conflict.

210 It should be acknowledged that the rather artificial task and setting certainly limits the external validity of
211 the findings. The usage of video footage, as well as requiring participants to simply press a button to indicate
212 behavioural intent is not fully comparable to a turning manoeuvre in real traffic. A confirmation of our
213 findings in a more realistic setting, e.g. on a test track, with actual cyclists approaching (Petzoldt et al.,
214 2017b), or at least in a driving simulator with virtual cyclists, is required to allow for more definite state-
215 ments on the potential effects of measures to improve e-bikes' safety. Also, the fact that the videos were of
216 variable length meant that the initial gap size differed between speed levels (from 10.3 s at 35 km/h to 24.0 s
217 at 15 km/h), which, at least in theory, might have constrained participants' indication of acceptable gaps.

218 Furthermore, from a practical point of view, to achieve a unique appearance, a re-design of the e-bike would
219 be preferable to a "re-design" of the rider. Any potential measure designed to improve the safety of e-bike
220 riders (and other road users who interact with them) should be independent of the motivation (or the
221 memory) of the rider. Our rather simple experimental implementation of a potential measure was only a
222 crude first attempt, with a number of potential flaws that would need to be addressed for a proper solution.
223 For example, one potential side effect of our specific implementation might be the fact that, while it would
224 result in e-bike riders being more visible in general, potentially improving their safety, it also carries the
225 potential for negative effects on the safety of other bicyclists (who would be less conspicuous in compari-
226 son). Furthermore, while our participants were not naïve with regard to the meaning of the orange helmet,

227 we cannot completely rule out the idea that, on a subconscious level, the selected colour might have been
228 associated with “risk”, and could somehow have influenced participants’ judgment. Still, with all due
229 acknowledgement of the investigation’s limitations, the central finding remains valid: There is some po-
230 tential, with the help of rather simple means, to modify road users’ gap acceptance behaviour, so that e-
231 bike riders might not simply have to wait for other road users to get used to them to finally be safe on the
232 road.

233 **ACKNOWLEDGMENTS**

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235 The authors would like to thank Pia Färber for her help while data acquisition.

236 **REFERENCES**

- 237 Alexander J, Barham P, Black I. Factors influencing the probability of an incident at a junction: results
238 from an interactive driving simulator. *Accident Analysis and Prevention*. 2002;34(6): 779–792.
- 239 Cavallo V, Laurent M. Visual information and skill level in time-to-collision estimation. *Perception*.
240 1988;17(5):623-632. doi:10.1068/p170623
- 241 Confederation of European Bicycle Industry [CONEBI]. 2017 Edition Industry & Market Profile. Brussels;
242 2017. Available at: <https://www.raivereniging.nl/ecm/?id=workspace://SpacesStore/072e2902-93de-4eb5-99dd-25227982fcaa>.
- 244 Cooper DF, Storr PA, Wennell J. Traffic studies at T-junctions - The effect of speed on gap acceptance and
245 conflict rate. *Traffic Engineering & Control*. 1977;18(3):110–112
- 246 Davies GM. Estimating the speed of vehicles: the influence of stereotypes. *Psychol Crime Law*.
247 2009;15(4):293-312. doi:10.1080/10683160802203971
- 248 Davies GM, Patel D. The influence of car and driver stereotypes on attributions of vehicle speed, position
249 on the road and culpability in a road accident scenario. *Leg Criminol Psychol*. 2005;10(1):45-62.
250 doi:10.1348/135532504X15394
- 251 Dozza M, Piccinini GFB, Werneke J. Using naturalistic data to assess e-cyclist behavior. *Transp Res Part*
252 *F Traffic Psychol Behav*. 2016;41(Part B):217-226. doi:10.1016/j.trf.2015.04.003
- 253 Hausteijn S, Møller M. E-bike safety: Individual-level factors and incident characteristics. *J Transp Heal*.
254 2016;3(3):386-394. doi:10.1016/j.jth.2016.07.001
- 255 Huertas-Leyva P, Dozza M, Baldanzini N. Investigating cycling kinematics and braking maneuvers in the
256 real world: e-bikes make cyclists move faster, brake harder, and experience new conflicts. *Transp Res Part*
257 *F Traffic Psychol Behav*. 2018;54:211-222. doi:10.1016/j.trf.2018.02.008
- 258 Jellinek R, Hildebrandt B, Pfaffenbichler P, Lemmerer H. MERKUR - Auswirkungen der Entwicklung des
259 Marktes für E-Fahrräder auf Risiken, Konflikte und Unfälle auf Radinfrastrukturen. Wien: Bundesministe-
260 rium für Verkehr, Innovation und Technologie; 2013
- 261 Keskinen E, Hiro O, Katila A. Older drivers fail in intersections: Speed discrepancies between older and
262 younger male drivers. *Accident Analysis and Prevention*. 1998;30(3):323–330.
263 [https://doi.org/10.1016/S0001-4575\(97\)00113-9](https://doi.org/10.1016/S0001-4575(97)00113-9)
- 264 McLeod RW, Ross HE. Optic-flow and cognitive factors in time-to-collision estimates. *Perception*.
265 1983;12(4):417-423. doi:10.1068/p120417
- 266 Petzoldt T, Schleinitz K, Heilmann S, Gehlert T. Traffic conflicts and their contextual factors when riding
267 conventional vs. electric bicycles. *Transp Res Part F Traffic Psychol Behav*. 2017a;46:477-490.
268 doi:10.1016/j.trf.2016.06.010
- 269 Petzoldt T, Schleinitz K, Krems J F, Gehlert T. Drivers’ gap acceptance in front of approaching bicycles –
270 Effects of bicycle speed and bicycle type. *Saf Sci*. 2017b;92:283-289. doi:10.1016/j.ssci.2015.07.021

- 271 Popovich N, Gordon E, Shao Z, Xing Y, Wang Y, Handy S. Experiences of electric bicycle users in the
272 Sacramento, California area. *Travel Behav Soc.* 2014;1(2):37-44. doi:10.1016/j.tbs.2013.10.006
- 273 Rose G. E-bikes and urban transportation: emerging issues and unresolved questions. *Transportation.*
274 2012;39:81-96. doi:10.1007/s11116-011-9328-y
- 275 Scaramuzza G, Uhr A, Niemann S. E-Bikes Im Strassenverkehr – Sicherheitsanalyse. Bern: bfu-Beratungs-
276 stelle für Unfallverhütung; 2015
- 277 Schleinitz K, Franke-Bartholdt L, Petzoldt T, Schwanitz S, Kühn M, Gehlert T. Pedelec-Naturalistic Cyc-
278 cling Study. Berlin: Unfallforschung der Versicherer e.V.; 2014
- 279 Schleinitz K, Petzoldt T. Can information about an approaching bicycle's characteristics influence drivers'
280 gap acceptance and TTA estimates? Paper presented at: 9th International Driving Symposium on Human
281 Factors in Driver Assessment, Training, and Vehicle Design.; June 25-28 2017:37-43. Manchester, Ver-
282 mont.
- 283 Schleinitz K, Petzoldt T, Franke-Bartholdt L, Krems J, Gehlert T. The German Naturalistic Cycling Study
284 - Comparing cycling speed of riders of different e-bikes and conventional bicycles. *Saf Sci.* 2017;92:290-
285 297. doi:10.1016/j.ssci.2015.07.027
- 286 Schleinitz K, Petzoldt T, Gehlert T. Drivers' gap acceptance and TTA judgements when confronted with
287 approaching bicycles, e-bikes and scooters. Paper presented at: International Cycling Safety Conference
288 2016. November 1-3 2016a; Bologna, Italy.
- 289 Schleinitz K, Petzoldt T, Krems JF, Gehlert T. The influence of speed, cyclists' age, pedaling frequency,
290 and observer age on observers' time to arrival judgments of approaching bicycles and e-bikes. *Accid Anal*
291 *Prev.* 2016b;92:113-121. doi:10.1016/j.aap.2016.03.020
- 292 Vlakveld WP, Twisk D, Christoph M, et al. Speed choice and mental workload of elderly cyclists on e-
293 bikes in simple and complex traffic situations: A field experiment. *Accid Anal Prev.* 2015;74:97-106.
294 doi:10.1016/j.aap.2014.10.018
- 295
- 296

297 **TABLES AND FIGURES**

298 **Table 1. Summary of ANOVA results for the gap size in seconds. Significant effects in boldface (N =**
 299 **49).**

	<i>df</i>	<i>F</i>	<i>p</i>	η^2_p
speed *	4	129.67	<.001	.73
bicycle type	1	0.06	.810	.00
helmet colour	1	21.67	<.001	.31
speed x bicycle type *	4	1.58	.199	.03
speed x helmet colour *	4	3.05	.040	.06
bicycle type x helmet colour	1	0.16	.689	.00
speed x bicycle type x helmet colour *	4	1.88	.141	.04

300 * Greenhouse Geisser correction

301

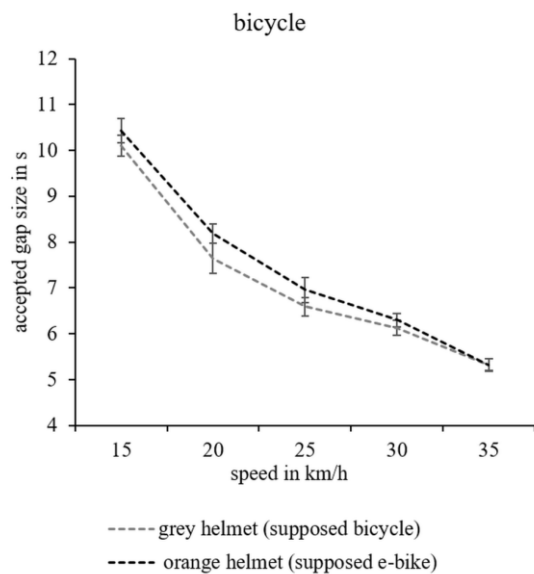
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303

304 **Figure 1. Screenshots from the video sequences (i.e., the observer's perspective). The horizontal**
 305 **white line marked the position of a potential collision between the oncoming cyclist and the ob-**
 306 **server (when turning left), rider wearing orange helmet (supposed e-bike).**

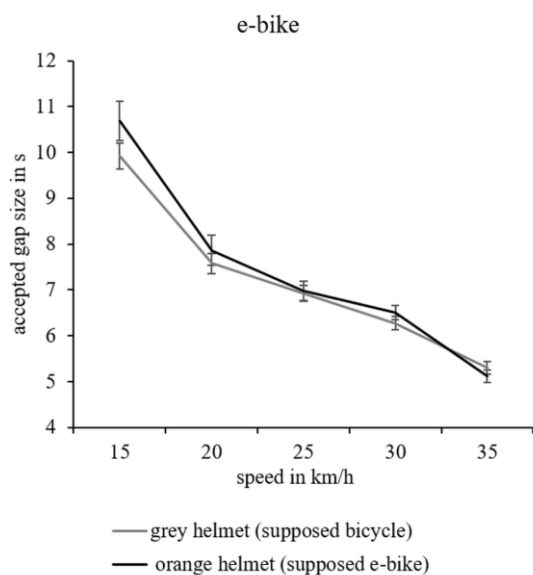
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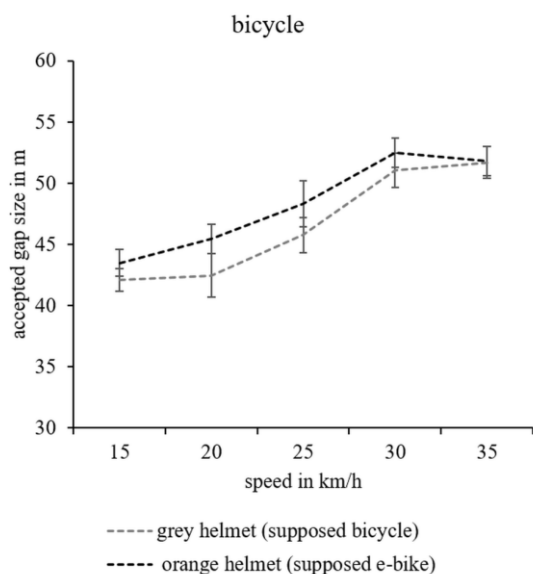
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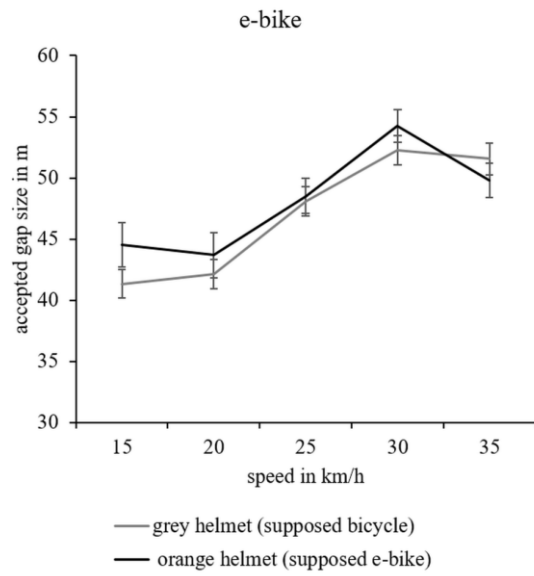
310 **Figure 2. Mean accepted gap size in seconds for the different speed levels dependent on helmet col-**
311 **our for the actual bicycle. Error bars represent 95% confidence intervals ($N = 49$).**
312



313 **Figure 3. Mean accepted gap size in seconds for the different speed levels dependent on helmet col-**
314 **our for the actual e-bike. Error bars represent 95% confidence intervals ($N = 49$).**
315
316



317 **Figure 4. Mean accepted gap size in meters for the different speed levels dependent on helmet col-**
318 **our for the actual bicycle. Error bars represent 95% confidence intervals ($N = 49$).**
319



320

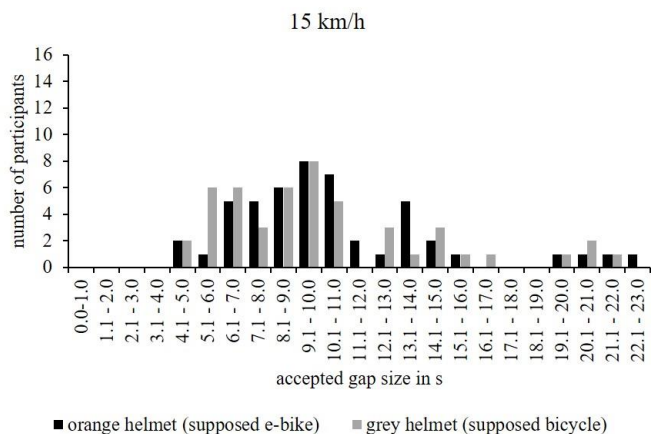
321

Figure 5. Mean accepted gap size in meters for the different speed levels dependent on helmet colour for the actual e-bike. Error bars represent 95% confidence intervals ($N = 49$).

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323

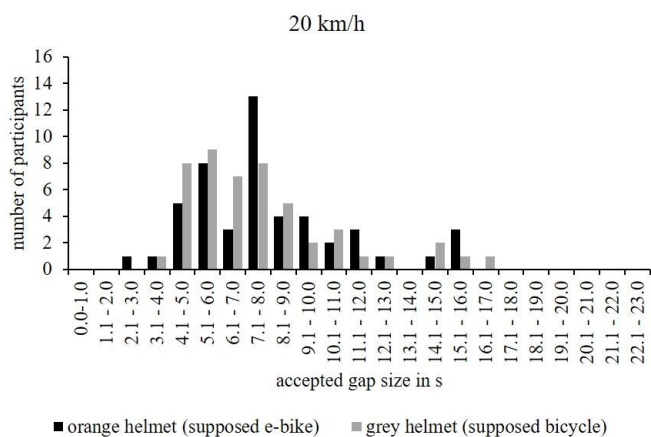
324 **APPENDIX**



325

326 **Figure A1. Distribution of responses for 15 km/h and the different helmet colours (N = 49).**

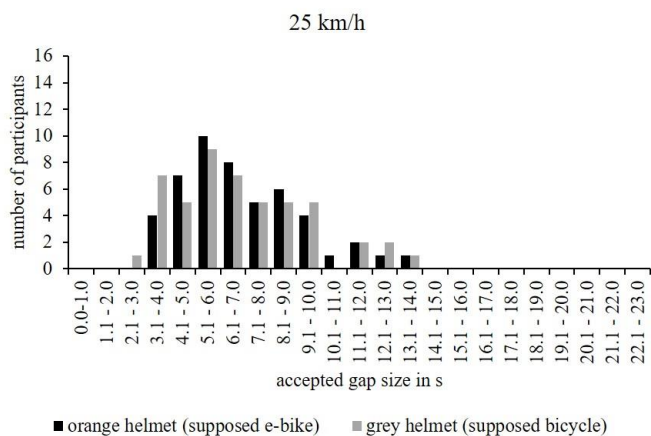
327



328

329 **Figure A2. Distribution of responses for 20 km/h and the different helmet colours (N = 49).**

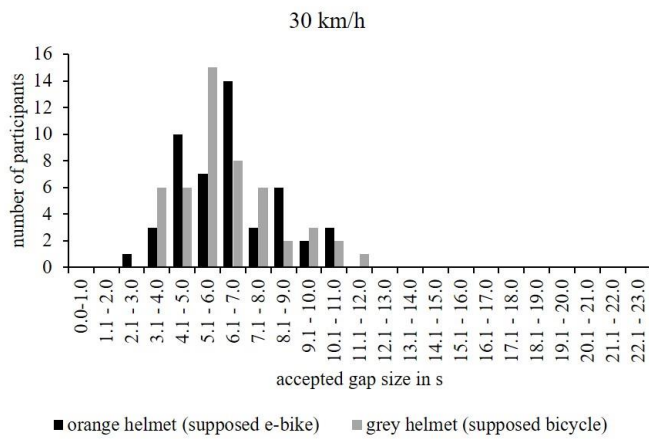
330



331

332 **Figure A3. Distribution of responses for 25 km/h and the different helmet colours (N = 49).**

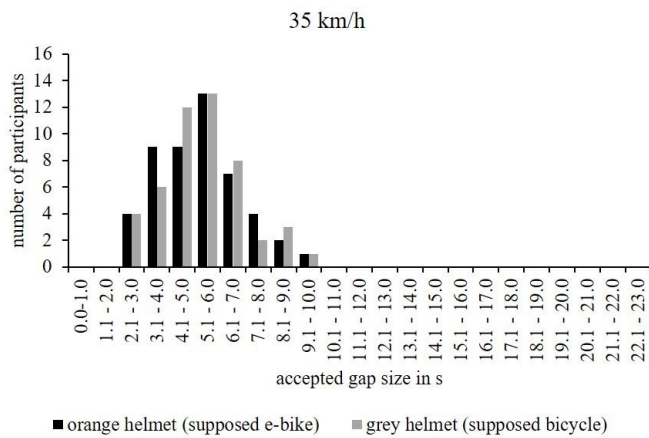
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334

335 **Figure A4. Distribution of responses for 30 km/h and the different helmet colours ($N = 49$).**

336



337

338 **Figure A5. Distribution of responses for 35 km/h and the different helmet colours ($N = 49$).**

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