

Exploring Age and Gender Differences in ICT Cybersecurity Behaviour

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Known age differences exist in relation to Information and Communication Technology (ICT) use, attitudes, access and literacy. Less is known about age differences in relation to cybersecurity risks and associated cybersecurity behaviours. Using an online survey, this study analyses data from 576 participants to investigate age differences across four key cybersecurity behaviours: Device securement, password generation, proactive checking, and software updating. Significant age differences were found; however, this is not a straightforward relationship. Older users appear less likely to secure their devices compared to younger users, however the reverse was found for the other behaviours, with older users appearing more likely to generate secure passwords, show proactive risk awareness and regularly install updates. Gender was not a significant predictor of security behaviour (although males scored higher for self-reported computer self-efficacy and general resilience). Self-efficacy was identified as a mediator between age and three of the cybersecurity behaviours (password generation, proactive checking and updating). General resilience was also a significant mediator for device securement, password generation and updating, however resilience acted as a moderator for pro-active checking. These findings have implications for the design of targeted training and development of interventions for different cybersecurity behaviours.

1. Introduction

More people are using digital technology than ever before, however 'digital divides' remain prevalent across user groups [1–3]. Demographic factors such as age and gender have often been cited as moderators of these digital divides. Younger age ranges have traditionally been the earliest adopters of ICT, however these age groups are reaching saturation (99% of young adults now use the internet in the UK [4]). Consequently, older adults are now the fastest growing group of adopters [4–6]. Despite many older adults being keen to adopt technology [7], a negative narrative prevails [8]. For example research suggests that this user-group may still lack confidence in their ability (or self-efficacy) to use their devices [9–11] and may show deficits in ICT skills and literacy [2,12,13], something often referred to as the 'second level' of the digital divide (where *access* to ICT forms the first level [1]). However, some researchers have argued that, rather than there being an age-related skills gap, older adults may simply underestimate their actual capabilities and knowledge [14]. In their review of this issue, Hunsaker and Hargittai [2] note a methodological issue with researching older adults, pointing out that studies differ in how they group age categories, the categories included and the age that is used to signal the start of older adulthood. They called for further work to identify whether age disparities are continuing.

For all users, the cost of embracing digital connectivity is a growing cybersecurity risk. As older adults now spend longer online, they in turn have become the latest target population for cyber-attacks, with £4m lost by older adults in the UK between 2018-19 [15]. However, research into age related differences in cybersecurity posture and attitude is scarce [11,16], which means it is difficult to identify and mitigate age-specific issues.

The risks that individual users are susceptible to may vary with age, but this is by no means conclusive. For example, while [17] suggests that younger users are more vulnerable to phishing

46 attacks, Grilli et al. [18] found that older adults were worse at discriminating between genuine and
47 phishing emails based on perceived suspiciousness. Sarno et al. [19] found no age differences in the
48 ability to classify emails as phishing or not. Oliveri et al. [20] discovered that older and younger
49 adults fall for different persuasion triggers, with older women being the most vulnerable group.
50 Other research suggests that younger adults display fewer privacy and security concerns compared
51 to older users (the latter potentially due to high levels of social media use and the associated sharing
52 of personal data [21,22]). Note, however, that this may not be a simple linear relationship, given a
53 study by Little and colleagues, who found a more complex U-shaped trend with younger and older
54 internet users appearing less protective of their privacy than their middle-aged counterparts [23].
55 Older adults show a reluctance to fully engage with cybersecurity behaviours, citing reasons
56 including low self-efficacy and a lack of awareness [11]. They are also less likely to adopt security
57 measures to protect against unauthorised access to their devices, e.g., PIN or biometric protections
58 [24]. Taken as a whole, the current research suggests that cybersecurity concerns may be more
59 complicated than simply identifying a single age range as vulnerable or 'at risk'. It is important that
60 we understand how adults of different ages engage with different security behaviours to protect
61 themselves online. This study addresses this gap in the literature and concentrates on four key
62 cybersecurity related behaviours: Device securement (e.g., locking their device screen when not in
63 use), secure password generation, proactive checking (checking legitimacy and security indicators
64 such as URLs and senders before clicking), and regular software updating.

65
66 Using data from across the adult lifespan (18-82yrs), the current study addresses some of the
67 limitations of previous research, where quite limited age ranges have been investigated (often due
68 to practical difficulties in data collection [25]). For example, Ayyagari and Crowell [26] recently
69 investigated differences between three age groups in relation to cybersecurity behaviours however
70 they were restricted to a university sample and their eldest group constituted anyone over 35 years.
71 In addition to assessing reported behaviours, we also expand the current literature by exploring the
72 role of computer self-efficacy, as this has been shown to influence ICT behaviour [27,28].
73 Psychological resilience has also been linked to risky behaviour. Specifically, resilience has been
74 linked to both risk seeking and risk adverse behaviours, depending upon the study and/or context
75 [29,30]. We therefore include a general resilience measure as a variable within our study.

76
77 This study also investigates gender differences as existing research in this area is inconclusive.
78 Traditionally research has suggested that females score lower for computer self-efficacy than males
79 [20,21] although more recently [22] suggest that this gender difference may be diminishing. It is
80 important to note that self-efficacy relates to the individuals own beliefs about how they can
81 perform [23]. As such, it is not possible to determine whether any gender differences reflect
82 differences in actual ability and/or differences in self-perception [24]. Computer self-efficacy can
83 also be context dependent, with several studies showing that gender differences may differ
84 depending on the context (e.g., ICT for educational versus general use; [25] or the specific task (e.g.,
85 internet tasks versus high level software-related tasks, [31]). Interestingly, some studies looking
86 specifically at cybersecurity behaviours report that females tend to show greater online privacy
87 concerns [27] and greater security policy compliance [28]. Whilst other studies show no gender
88 differences, for example Vance and colleagues [32] found no gender differences for intention to
89 comply with security policies, and others suggest that females are likely to act less securely [33]. In
90 their review of older adult research, Hunsaker and Hargittai also described the existing literature as
91 inconclusive. We address this need for increased understanding by including gender analyses in the
92 current study.

93 In summary, our study tests for age and gender differences in cybersecurity behaviour across the
 94 adult lifespan, after controlling for computer self-efficacy and general resilience. The results have
 95 implications for identifying priority areas for future targeted training and development
 96 interventions.

97 **2. Materials and Methods**

98 Full ethical approval was granted from the School of Health and Life Sciences ethics committee at
 99 Northumbria University. An online survey was distributed by online recruitment platform
 100 'Prolific.ac'. Prolific is a paid service that distributes online questionnaires to their userbase of
 101 participants. The initial sample of 607 responses was cleaned and 31 responses removed due to
 102 failing the 'attention check' question. The final sample consists of data from 579 participants, aged
 103 18-82 years ($M = 33.86\text{yrs}$, $SD = 11.80\text{yrs}$). Further demographics are shown in Table 1.

104
 105 Table 1. Sample demographics (N = 579)

		N	%
Age	18-24	131	22.6
	25-34	219	37.8
	35-44	143	24.7
	45-54	46	7.9
	55-64	26	4.5
	65-74	12	2.1
	75-82	2	0.3
Gender	Male	236	40.8
	Female	340	58.7
	Other	3	0.5
Education	Primary/Elementary School	4	.07
	Secondary/High School	67	11.6
	College/A-Level	146	25.2
	Bachelors	239	41.3
	Masters	98	16.9
	Doctorate	25	4.3
Country	UK	275	47.5
	USA	152	26.2
	Canada	152	26.3

106
 107 In addition to the demographic questions, participants were asked to complete a series of scale
 108 items to measure their cybersecurity behaviour, their computer self-efficacy and their general
 109 resilience. Cybersecurity related behaviour was measured using the Security Behaviour Intentions
 110 Scale (SeBIS) [34]. SeBIS is a 16-item scale consisting of four subscales that measure attitudes
 111 towards device securement, password generation, proactive checking and software updating. The
 112 scale showed acceptable reliability in our study with Cronbach's alpha (α) ranging from .64-.75 for
 113 the four subscales (see Table 3). The Computer Self-Efficacy scale [35] was used to measure users'
 114 beliefs about their ICT capabilities. The scale showed excellent reliability ($\alpha = .93$). General resilience
 115 was measured using the Brief Resilience Scale [36] ($\alpha = .89$).

116 **3. Results**

117 Missing data accounted for less than 0.3% of the items. Little’s MCAR test was non-significant (χ^2
 118 (117) = 118.88, $p = .43$) indicating that the data was missing completely at random, therefore
 119 Estimated Maximum Likelihood was used to compute the missing data. Due to insufficient sample
 120 size ($n=3$) the other gender category was excluded from the analyses.

121
 122 Data was checked to ensure it met the assumptions of normality, independence and
 123 homoscedasticity. All values were checked to ensure that they were within the expected ranges
 124 given the measurement scales used. There was no sign of multicollinearity between the predictor
 125 variables (all correlations $<.7$, see Table 2; VIF scores <2), scatterplots indicated a linear relationship
 126 between the IVs and DVs and plotting the standardised residuals and predicted values indicated
 127 adequate homoscedasticity. All dependent variables appeared normally distributed on the Q-Q plots
 128 (and skew and kurtosis values <2), except for device securement. The latter indicated negative skew
 129 (more scores towards the top of the scale) although this was still within the acceptable threshold of
 130 ± 2 [37]. Device securement also showed a kurtosis value of 2.28. Therefore, as the normality
 131 assumption was violated for device securement, all analyses using this variable were conducted
 132 using the bootstrapping method (with bias-corrected and accelerated confidence intervals, samples
 133 = 2000) to ensure robustness.

134
 135 Bivariate correlations are shown for each of the variables (Table 2). There is no significant correlation
 136 between age and gender. None of the correlations raise concerns around multicollinearity.

137
 138 Table 2. Bivariate correlations for each of the variables

	M	SD	1.	2.	3.	4.	5.	6.	7.	8.
1. Age	33.85	11.82	-							
2. Gender	-	-	.03	-						
3. S.Efficacy	3.64	.82	-.11**	.39***	-					
4. Resilience	3.24	.85	.11**	-.12**	.23***	-				
5. Device Securement	4.06	.83	-	-.01	.10*	.11**	-			
6. Password Generation	3.29	.82	.18***	.15***	-.17***	.27***	.21***	.20***	-	
7. Proactive Checking	3.71	.69	.12**	-.16***	.31***	.17***	.19***	.46***	-	
8. Updating	3.42	.87	.14***	-.17***	.33***	.20***	.18***	.32***	.29***	-

139 Note: *** $p < .001$ ** $p < .01$ * $p < .05$

140 **3.1. Age and Gender differences in perceived Computer Self-Efficacy and General Resilience**

141 Independent samples t-tests showed a significant difference between the genders, with males ($M =$
 142 4.02 , $SD = .74$) scoring significantly higher than females for perceived computer self-efficacy ($M =$
 143 3.38 , $SD = .78$), $t(574) = 9.99$, $p < .001$. T-tests also show a significant difference between the genders
 144 for general resilience, with males ($M = 3.21$, $SD = .81$) scoring significantly higher than females ($M =$
 145 3.16 , $SD = .88$), $t(574) = 2.87$, $p = .004$. There were no significant age differences for self-efficacy or
 146 resilience scores.

147 **3.2. Predictors of Cybersecurity Behaviours**

148 The data were analysed using a series of hierarchical regressions to test the predictors (age, gender,
149 computer self-efficacy and general resilience) of cybersecurity behaviour. As aforementioned, the
150 device securement regression was conducted using the bootstrapping method due to violating the
151 assumptions of homoscedasticity, therefore confidence intervals are reported for this regression.
152 All four models were significant (Table 3): Device securement (Bootstrap Samples = 2000, $R^2 = .05$,
153 BCa CI [.03 - .08]), Password generation ($F(4,571) = 12.06, p < .001, R^2 = .13$), Proactive checking
154 ($F(4,571) = 20.19, p < .001, R^2 = .12$), and Updating ($F(4,571) = 25.13, p < .001, R^2 = .15$).

155
156 Investigating the individual predictors revealed that age was a significant predictor for all four
157 cybersecurity behaviours (Table 3). Age was a negative predictor of device securement, but a
158 positive predictor for the other behaviours (password generation, proactive checking and updating).
159 Gender was not a significant predictor for any of the behaviours.

160
161 The standardised coefficients show the strongest predictors. For three of the four behaviours
162 (password generation, proactive checking and updating), computer self-efficacy was the strongest
163 predictor, followed by age and then general resilience. All of which were positive predictors.

164
165 Device securement differed from the other behaviours. The strongest predictor variable, age, acted
166 as a negative predictor of this behaviour. General resilience was the only other significant predictor,
167 acting as a positive predictor of secure behaviour.

168

169 Table 3. Regression results

	Device Securement [±] (α=.64)					Password Generation (α=.71)					Proactive Checking (α=.66)					Updating (α=.75)									
	B	Beta	SE	BCa CI	R ² [CI]	B	Beta	SE	<i>t</i>	R ²	B	Beta	SE	<i>t</i>	R ²	B	Beta	SE	<i>t</i>	R ²					
Age	-.01	-.18	.00	-.02, -.01***		.01	.16	.00	4.09***		.01	.14	.00	3.51***		.01	.16	.00	4.16***						
Gender	.07	.04	.08	-.09, .22		-.11	-.07	.07	-1.61		-.06	-.04	.06	-.92		-.08	-.05	.07	-1.12						
S.Efficacy	.07	.07	.05	-.02, .16		.23	.23	.04	5.38***		.24	.29	.04	6.58***		.32	.30	.05	7.01***						
Resilience	.12	.12	.05	.03, .21*	.05	.13	.13	.04	3.31**	.13*	.07	.08	.03	2.02*	.124**	.11	.11	.04	2.61**	.15*					
	[.03, .08]																								

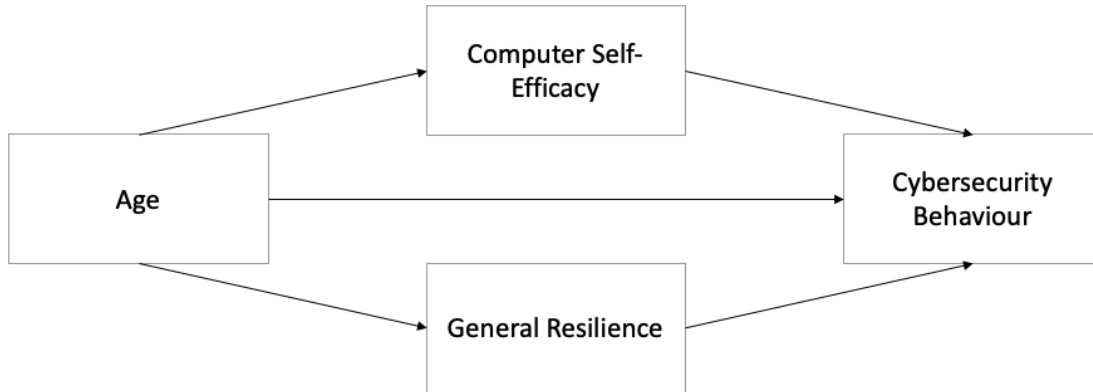
170 Note: ±2000 bootstrap samples. ****p*<.001 ***p*<.01 **p*<.05

171

172

173 **3.3. Mediation Analysis**

174 The relationship between age and perceived computer self-efficacy and resilience was investigated
175 further with parallel mediation analysis using the PROCESS macro for SPSS, model 4 (Hayes, 2013,
176 Figure 1).
177



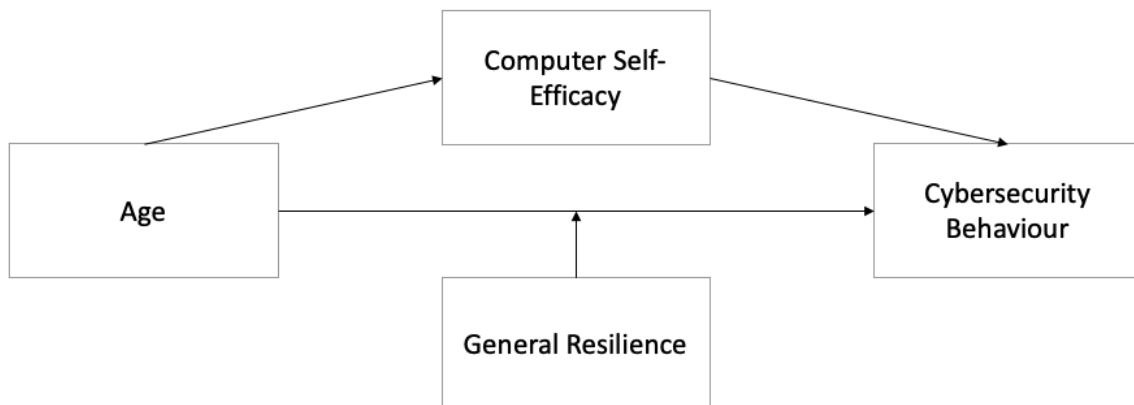
178
179 Figure 1. Parallel mediation model (PROCESS model 4)

180
181 To aid interpretation of the results, all variables that defined products were mean centered during
182 the PROCESS mediation analysis. The results are shown in Table 4.

183
184 The indirect effect of age on cybersecurity behaviour, via self-efficacy (mediator 1), was significant
185 for three of the four behaviours: Password generation, proactive checking and updating. Self-efficacy
186 was not a significant mediator for device securement.

187
188 The indirect effect of age on cybersecurity behaviour, via resilience (mediator 2), was significant for
189 three of the four behaviours: Device securement, password generation and updating. The effect of
190 resilience on the remaining cybersecurity behaviour, proactive checking, was investigated using
191 PROCESS model 5. The results indicate that for this behaviour, resilience acts as a moderator rather
192 than a mediator. The tested model is shown in Figure 2.

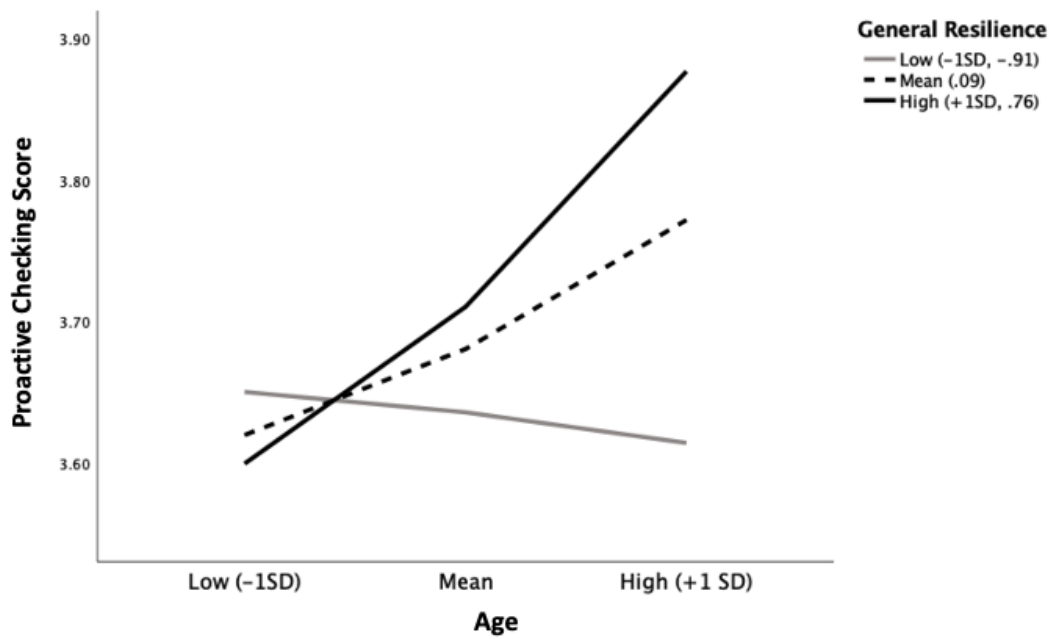
193



194
195 Figure 2. Model testing mediation via computer self-efficacy and moderation via general resilience
196 (PROCESS model 5)

197
198
199
200
201

Plotting the estimates shows that the moderation effect of resilience on proactive checking for low (-1SD), mean, and high (+1SD) age (Figure 3). The effect of age on proactive checking is strongest for the high resilience users.



202
203
204

Figure 3. Moderation of the effect of general resilience on proactive checking across low, mean and high age groups.

205 Table 4. Parallel mediation analysis for each of the cybersecurity behaviours. Significant paths are indicated with an asterix(*).

	Device Securement			Password Generation			Proactive Checking			Updating		
	<i>B</i>	<i>SE</i>	CI	<i>B</i>	<i>SE</i>	CI	<i>B</i>	<i>SE</i>	CI	<i>B</i>	<i>SE</i>	CI
Total Effect of Age	-.01	.00	[-.02, -.01]*	.01	.00	[.00, .02]*	.01	.00	[.00, .01]*	.01	.00	[.00, .02]*
Direct Effect of Age	-.01	.00	[-.02, -.01]*	.01	.00	[.01, .02]*	.01	.00	[.00, .01]*	.01	.00	[.01, .02]*
Indirect Effect via Self-Efficacy	-.00	.00	[-.00, .00]	-.00	.00	[-.00, -.00]*	-.00	.00	[-.00, -.00]*	-.00	.00	[-.00, -.00]*
Indirect Effect via Resilience	.00	.00	[.00, .00]*	.00	.00	[.00, .00]*	.00	.00	[.00, .00]	.00	.00	[.00, .00]*

206

207

208 4. Discussion

209 This study expands upon the current literature by investigating age and gender differences in
210 relation to different cybersecurity behaviours. Our results show that rather than older adults being
211 universally more at risk than others, age differences vary according to the specific security behaviour
212 in question. Therefore, rather than focusing on first level digital divides (i.e., ICT access and
213 adoption) our findings highlight the importance of investigating ICT behaviour on a more granular
214 level, i.e., investigating specific types of behaviour and/or activities (something also identified by
215 [19]). Whilst younger users appear more likely to secure access to their devices than the older age
216 groups, they also appear *less* likely to generate secure passwords and/or update their device, and
217 show less proactive URL/email checking behaviours. Our result regarding proactive checking
218 provides a reason younger users may be more susceptible to phishing [17], and older adults to be
219 less likely to adopt security measures to secure physical use of their devices [24]. Similarly, a recent
220 study [38] found that – in direct contrast to their original hypothesis - older users are less likely to
221 share their passwords. Our study helps to strengthen the emerging positive discourse that older
222 users are security conscious, challenging dated stereotypes that this age group are not tech savvy
223 [14]. Many older adults actually display high levels of awareness and ability in regards to
224 cybersecurity [38–40].

225
226 It can appear contradictory that older users on the one hand are security conscious and generate
227 more secure passwords but are also *less* likely to secure access to their devices, e.g., failing to lock
228 their screen when the device is not in use. On further consideration, this may be due to differences
229 in perceived risks. Existing literature suggests that this age group focuses heavily upon the privacy
230 and security of the data they enter online [41] – which is in keeping with our results which show they
231 are more likely to generate strong passwords, update devices and show proactive checking for risk.
232 In comparison, it is possible that they are not as aware or not as cautious of ‘offline risks’ around the
233 security of their physical device, such as it being stolen or used maliciously. For instance, if their
234 main point of access is a home computer, they may feel they the device is already secure within the
235 home and that there is little risk of other people accessing it [11]. Interventions to increase the
236 salience and importance of physical device securement may be beneficial for this age group. Based
237 on the existing literature, the most favoured and/or effective intervention approaches for older
238 adults may be those involving in-person support and/or promoting these security behaviours
239 through social connections, peer support and family members [3,40,41,43]. However, it is also
240 important to note that a lack of device securement may be an active choice on behalf of some users
241 and may not represent a lack of awareness. For example, it is possible that older adults knowingly
242 allow others to access their devices, for example research suggests older adults may be more likely
243 to ask trusted others to complete ICT tasks on their behalf [11,44]. There may also be barriers due to
244 problems with biometric security, for example Morrison and colleagues identified that fingerprint
245 readers can be problematic for older users [11].

246
247 Similarly, if younger users are the earliest and most intensive users of ICT, and they are *more* likely to
248 secure access to their devices, why is it that they appear to be *less* likely to generate secure
249 passwords, demonstrate pro-active checking for risk or update their devices? Some of these findings
250 could potentially be explained through differences in usage and/or device type. For example, in
251 relation to secure passwords – it could be that younger users are relying more heavily upon
252 automatic password generators [11,45] and/or biometrics (e.g., face ID, fingerprints), therefore
253 removing much of the emphasis on personally generating a secure password. In relation to proactive

254 checking for risk, frequent ICT usage and over familiarisation with the sharing of personal data can
255 lead to overconfidence, complacency and/or security fatigue [46–49] – factors which have been
256 linked to cybersecurity vulnerability [50]. It also possible that the salience of a possible attack may
257 be reduced in the younger age groups due to a lack of learned experience (i.e., not having personally
258 suffered an attack, or heard of friends or family being affected). Regarding younger users reporting
259 being less likely to update their devices, many devices now automatically install software updates as
260 they become available. Trust in automation could lead to users feeling less responsibility and reduce
261 the requirement to check whether their devices are up to date. Our mediation results also suggest
262 that self-efficacy is a significant mediator of age and security behaviour, therefore suggesting that, at
263 least to some extent (and again potentially related to a reliance upon automation), younger users
264 may demonstrate reduced self-efficacy compared to older users. Further qualitative and quantitative
265 research is necessary to identify the factors underlying the age differences and the role of efficacy
266 identified in this study. These insights can help to guide the design of future interventions to
267 promote more secure behaviour.

268

269 It is not unexpected that computer self-efficacy would positively predict some cybersecurity
270 behaviour given that it relates to the individual's confidence in their IT capabilities (a similar result
271 was found by Mitzer et al. [51]) and therefore their ability to act securely. It is perhaps more
272 surprising that general resilience was a significant positive predictor across all four behaviours. It
273 could have been expected that resilience would act as a negative predictor due to being associated
274 with self-confidence in 'bouncing back' if anything bad happens, and therefore perhaps less
275 incentive to avoid risks. However, the literature shows that the relationship between resilience and
276 risk is not this simple. It has been suggested that resilience negatively predicts negative health
277 behaviours (e.g., smoking, heavy drinking, drug use) and positively predicts protective health and
278 safety promotion behaviours (e.g., wearing a seatbelt, eating a healthy diet, exercising, crossing the
279 street safely) [29]. This resonates with our results as the behaviours we were predicting were safety
280 promoting. Our findings indicate the general resilience acts as a mediator for three of the four
281 behaviours (device securement, password generation and updating) and as a moderator for the
282 remaining behaviour, pro-active checking for risk. The greatest effect of age on pro-active checking
283 was found for those users who scored high for general resilience. One potential explanation is that
284 younger users' perceptions of resilience may be based more on optimism bias (i.e., feeling resilient
285 but not being proactive to protect against risk), whereas older users' resilience may be based more
286 upon learned experience (and therefore their learned abilities to act proactively to protect against
287 risk in the future). Future research may wish to further investigate the role of resilience in relation to
288 online behaviour.

289

290 Interestingly, we found no evidence of gender differences in relation to any of the cybersecurity
291 behaviours. There was a gender difference for computer self-efficacy scores; with males scoring
292 significantly higher than females. This is not unexpected as this trend has traditionally been reported
293 in the previous literature [52]. As self-efficacy can be context specific [31], it is also possible that the
294 Computer Self-Efficacy Scale [35] measures self-efficacy in relation to tasks that males generally feel
295 more confident with. Furthermore - and as noted earlier - self-efficacy relates to an individual's own
296 beliefs about their ability; and does not necessarily reflect actual differences in ability or
297 performance [53]. Even so, it is worth noting, that our findings are contrary to research suggesting
298 that gender differences in perceptions of computer self-efficacy may have abated in recent years
299 [54]. We also found that males scored significantly higher on general resilience, this is a trend that
300 has been observed in the existing literature [55]. Previous research [56] has attributed higher male

301 resilience scores to differences in self-perception and cultural constructions of ‘masculinity’.

302

303 We recognise the limitations within the current study and make recommendations for future
304 research. Firstly, whilst we included a broad range of ages, most of our participants were below 45
305 years of age. Future research should seek to follow the recommendations of Hunsaker and Hargittai
306 [2], who call for research to include more subcategories of older adults (see for example [57] who
307 use the categories 55-64yrs, 65-79yrs and 80-97yrs). With more granular analysis of older age
308 groups, it is possible that further group disparities and more complex relationships could emerge
309 (such as U-shaped trends similar to those found by [23].) Secondly, we recognise that this study
310 relies upon self-reported data, and we suggest that future research utilises experimental and/or
311 observational methods. Thirdly, our participants were recruited via an online recruitment platform,
312 therefore they may be more tech-savvy than the general population (similar to that found for mTurk
313 users, e.g., [17]). It should be recognised that they may not be representative of the larger
314 population of ICT users.

315 **5. Conclusions**

316 In this paper we identify behaviour-specific age differences in cybersecurity, highlighting the need
317 for a granular, context-specific approach to identify age-related differences in cybersecurity
318 behaviours; and advise against labelling a particular age group as universally more at risk. While
319 within our sample, older users were more likely to report generating secure passwords, updating
320 their devices and demonstrating pro-active checking for risk. In comparison, they were less likely to
321 secure their device to prevent unauthorised access (e.g., by locking the screen), the relationship
322 between age and security behaviour was mediated by computer efficacy for three of the four
323 behaviours, with the exception being device securement. This indicates that a lack of device
324 securement by older users is due to other reasons, this could include low perceived risk of physical
325 access to devices by malicious parties, and/or an active choice to allow access by others such as
326 family members. General resilience was also a mediator for three of the four behaviours, and a
327 moderator for the remaining behaviour (proactive checking for risk). The relationship between age
328 and pro-active checking was strongest for those users scoring high for resilience. We suggest that
329 this may represent a move from optimism bias in younger users to learned experience (and
330 therefore learned protective mechanisms) in older users. We present multiple recommendations for
331 future research to further explore the impact of age, self-efficacy and resilience on cybersecurity
332 behaviour. Despite gender differences in self-perceived computer self-efficacy and general
333 resilience, no gender differences were found for the cybersecurity behaviours; suggesting that
334 gender does not play a role in cybersecurity behaviour intentions. These findings have implications
335 for future design and development of targeted cybersecurity interventions.

336

337 **Data Availability**

338

339 The survey data used to support the findings of this study have been deposited in the University of
340 Bath data archive (DOI will be provided upon publication).

341

342 **Conflicts of Interest**

343

344 The authors declare no conflicts of interest.

345

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347

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