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# Exploring Age and Gender Differences in ICT Cybersecurity Behaviour

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# Exploring Age and Gender Differences in ICT Cybersecurity Behaviour

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)

		<b>N</b>	<b>%</b>
<b>Age</b>	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
<b>Gender</b>	Male	236	40.8
	Female	340	58.7
	Other	3	0.5
<b>Education</b>	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
<b>Country</b>	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 ( $\alpha$ ) 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 ( $\chi^2$   
 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  $\pm 2$  [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	.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 <sup>±</sup> (α=.64)					Password Generation (α=.71)					Proactive Checking (α=.66)					Updating (α=.75)									
	B	Beta	SE	BCa CI	R <sup>2</sup> [CI]	B	Beta	SE	<i>t</i>	R <sup>2</sup>	B	Beta	SE	<i>t</i>	R <sup>2</sup>	B	Beta	SE	<i>t</i>	R <sup>2</sup>					
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

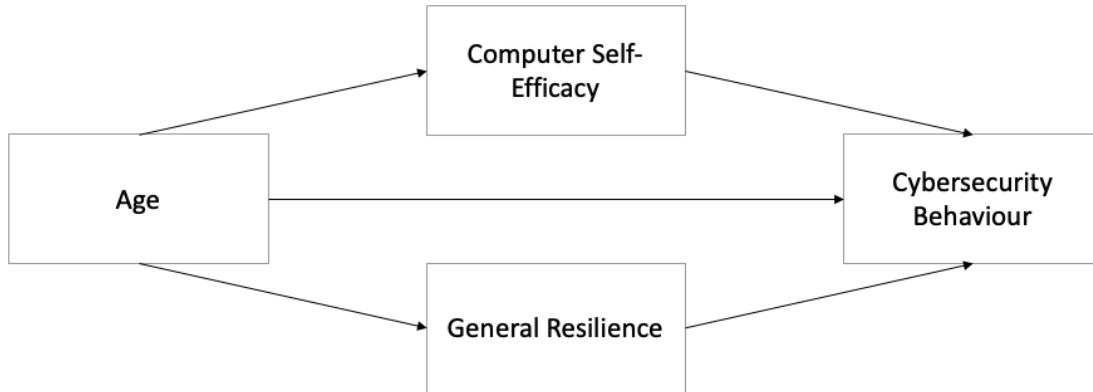
171

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



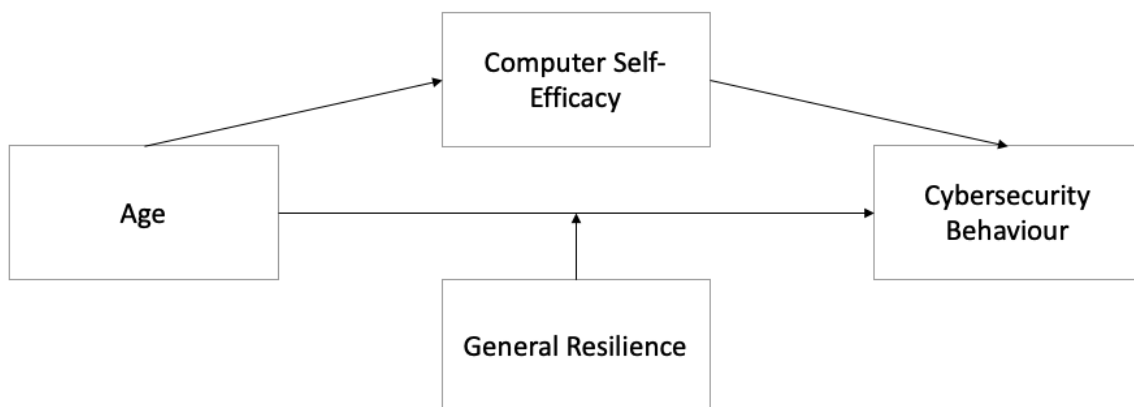
178  
179 Figure 1. Parallel mediation model (PROCESS model 4)

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.

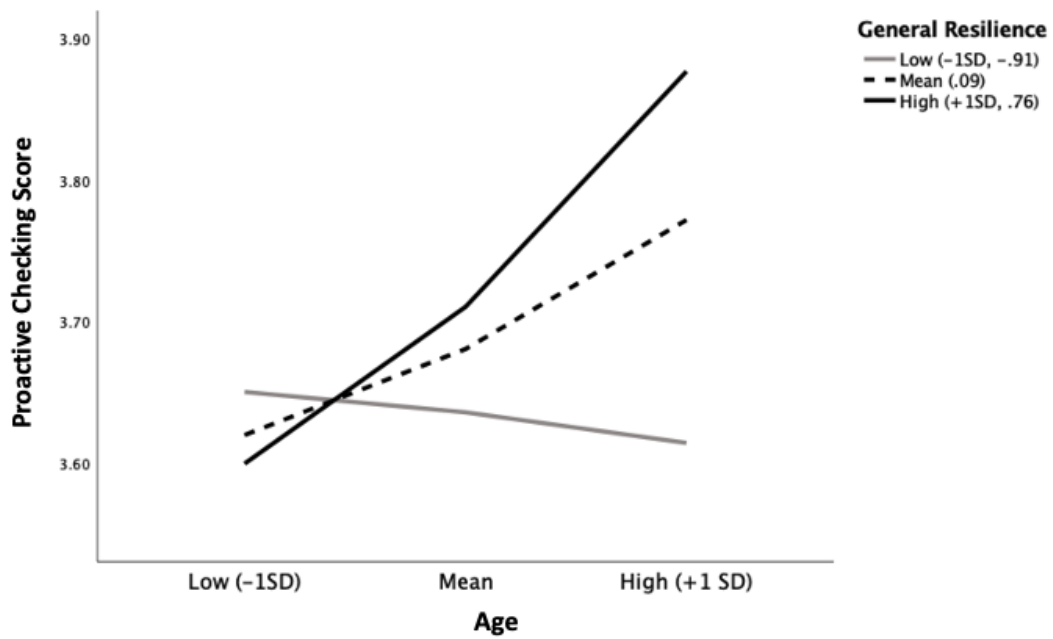
193



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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