

# Facelock: Familiarity-based graphical authentication

Authentication codes such as passwords and PIN numbers are widely used to control access to resources. One major drawback of these codes is that they are difficult to remember. Account holders are often faced with a choice between forgetting a code, which can be inconvenient, or writing it down, which compromises security. In two studies, we test a new knowledge-based authentication method that does not impose memory load on the user. Psychological research on face recognition has revealed an important distinction between familiar and unfamiliar face perception: When a face is familiar to the observer, it can be identified across a wide range of images. However, when the face is unfamiliar, generalisation across images is poor. This contrast can be used as the basis for a personalised 'facelock', in which authentication succeeds or fails based on image-invariant recognition of faces that are familiar to the account holder. In Study 1, account holders authenticated easily by detecting familiar targets among other faces (97.5% success rate), even after a one-year delay (86.1% success rate). Zero-acquaintance attackers were reduced to guessing (<1% success rate). Even personal attackers who knew the account holder well were rarely able to authenticate (6.6% success rate). In Study 2, we found that shoulder-surfing attacks by strangers could be defeated by presenting different photos of the same target faces in observed and attacked grids (1.9% success rate). Our findings suggest that the contrast between familiar and unfamiliar face recognition may be useful for developers of graphical authentication systems.

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**Facelock: Familiarity-based graphical authentication**

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15 **Introduction**

16 Security codes such as passwords and personal identity numbers (PINs) are  
17 widely used to control access to resources (e.g. bank accounts, websites, mobile  
18 devices). To protect against fraudulent access, it is essential that a security code  
19 should be difficult to guess (Garfinkel & Spafford, 1996; Gehringer, 2002; Carstens  
20 & Malone, 2009). From this standpoint, a random sequence of symbols (e.g.  
21 "8z3gxFtv") is a much better password than a user's own surname (e.g.  
22 "jenkins"). However, security codes that are difficult to guess tend also to be  
23 difficult to remember (Ebbinghaus, 1964; Craik & Lockhart, 1972; Zviran & Haga,  
24 1990; Zviran & Haga, 1993). For this reason, legitimate code holders are often  
25 faced with a choice between forgetting a code, which can be frustrating and  
26 inconvenient, or writing it down, which compromises security (Carstens, 2009;  
27 Tam, Glassman, & Vandenwauver, 2010).

28 These and other weaknesses (Adams & Sasse, 1999; Sasse, Brostoff, & Weirich,  
29 2001) have led developers to explore other forms of knowledge-based  
30 authentication, including graphical authentication (Blonder, 1996; Biddle,  
31 Chiasson, van Oorschot, 2012). In such systems, a user's authentication code is a  
32 set of images rather than an alphanumeric string. To log in, users identify their  
33 own images from larger challenge sets (Podd, Bunnell, & Henderson, 1996;  
34 Brostoff & Sasse, 2000; Dhamija & Perrig, 2000; Furnell, Papadopoulos, &  
35 Dowland, 2004; Weinshall & Kirkpatrick, 2004). One of the most well developed of  
36 these systems is Passfaces (Brostoff & Sasse, 2000), in which the images used are  
37 photographs of faces. Passfaces offers several advantages over standard  
38 passwords, most notably higher memorability of authentication codes (Paivio &  
39 Csapo, 1973). For example, an early evaluation found that after a 5-month delay,  
40 72% of participants remembered their Passfaces codes on their first login attempt  
41 (Valentine, 1998). For comparison, a similar evaluation of passwords found that  
42 only 27% of passwords were remembered following a delay of 3 months (Zviran &

43 Haga, 1993). In a pioneering field trial, Brostoff & Sasse (2000) reported that login  
44 failures were three times higher for passwords than for Passfaces. This estimate is  
45 consistent with previous findings. However, such graphical systems are not  
46 without their limitations (Furnell, Papadopoulos, & Dowland, 2004; Tari, Ozok, &  
47 Holden, 2006; Everitt, Bragin, Fogarty, & Kohno, 2009; Mihajlov & Jerman-Blazic,  
48 2011). Perhaps foremost among these is their susceptibility to 'shoulder-surfing'  
49 attacks (Tari, Ozok, & Holden, 2006), in which an attacker obtains a user's  
50 authentication code by secretly watching the user during authentication. This  
51 attack is powerful because it exploits the memorability of image-based codes:  
52 Images that are easy for the user to recognise are also easy for an attacker to  
53 recognise (Paivio & Csapo, 1973).

54 In the present study we show that this symmetry - between ease of recognition  
55 for the user, and ease of recognition for the attacker - can be broken by applying  
56 insights from cognitive psychology research. Psychological studies of face  
57 recognition have revealed strong qualitative differences between processing of  
58 familiar and unfamiliar faces (Burton & Jenkins, 2011; Jenkins & Burton, 2011).  
59 When a face is familiar to the viewer, it can be identified from a wide range of  
60 different photographs, even when image quality is very poor (Harmon, 1973;  
61 Burton, Wilson, Cowan, & Bruce, 1999; Burton, Jenkins, & Schweinberger, 2011;  
62 see Figure 1). Importantly for this study, different images of a familiar face are  
63 almost never mistaken for different people (Jenkins, White, van Montfort, &  
64 Burton, 2011). In contrast, our ability to identify unfamiliar faces from  
65 photographs is strikingly poor (Bruce et al., 1999, 2001). Very often, different  
66 photos of an unfamiliar face are seen as different individuals (Jenkins, White, van  
67 Montfort, & Burton, 2011). Thus, familiarity with a particular face determines  
68 one's ability to identify it across changes in image (see Figure 2). Although the  
69 transformative effect of familiarity on face recognition may be not be intuitively  
70 obvious, it is highly robust, and has been replicated in dozens of experiments

71 spanning decades of research (Bruce, 1982; Clutterbuck & Johnston, 2002, 2004,  
72 2005; Megreya & Burton, 2006; Jenkins, White, van Montfort, & Burton, 2011).

73 The familiarity contrast is normally encountered as a *problem* in applied settings.  
74 For example, unfamiliar face matching presents a serious challenge for security  
75 personnel and for automatic face recognition systems. In the present study we  
76 offer a very different perspective, by describing how the familiarity contrast might  
77 be exploited positively as the basis of an authentication system. The principle is  
78 straightforward: Familiarity with a particular face determines an observer's ability  
79 to identify it across different photographs. For any individual face that is not  
80 widely known, this ability will be very narrowly concentrated within the  
81 population. If a *set* of such faces is known only to a single individual, it can be used  
82 to create a personalized lock. Access is granted to anyone who demonstrates  
83 image-invariant recognition of the critical faces, that is, anyone who is familiar  
84 with them all. Conversely, access is denied to anyone who does not demonstrate  
85 image-invariant recognition of the critical faces, that is, anyone who is not familiar  
86 with them all.

87 To test this principle, we developed a prototype system that involves presenting a  
88 series of face arrays, similar to Passfaces. In our scheme, each array contains one  
89 face that is familiar to the user, among other faces that are unfamiliar. The user  
90 gains access by simply indicating the familiar face in each array. We refer to this  
91 method as Facelock. The scheme has two major advantages over traditional  
92 authentication methods. First, there is no explicit memory involved - the task is  
93 simply to pick out the familiar face in each array. As this task does not require the  
94 user to remember a code, the issue of forgetting one's code does not arise.

95 Dispensing with a set code also means that the challenge arrays, and the familiar  
96 faces embedded in them, may be composed of different photographs of different  
97 individuals at each login. This is very different from the traditional approach of  
98 assigning a single invariant authentication code to an account holder. The second

99 major advantage concerns the problem of writing down authentication codes  
100 (Dunphy, Nicholson, & Oliver, 2008; Tam, Glassman, & Vandenwauver, 2010). In  
101 the proposed scheme, the user is not required to reproduce a particular set code  
102 in order to authenticate. The only requirement is to distinguish familiar faces from  
103 unfamiliar faces. As familiarity discriminations are extremely robust (Young, Hay,  
104 & Ellis, 1985; De Haan, Young, & Newcombe, 1991), users have no incentive to  
105 write down aide-memoires for their targets, and the associated security risk can  
106 be avoided.

107 The main aims of the current work are i) to test the feasibility of an authentication  
108 method that exploits the familiarity contrast in face recognition, and ii) to assess  
109 its resilience against two very different forms of attack - guessing by high-  
110 acquaintance attackers, and shoulder-surfing by zero-acquaintance attackers. The  
111 aim at this stage is not to develop a commercially viable system. Instead we seek  
112 to raise awareness of the important psychological contrast between familiar and  
113 unfamiliar face processing, and to explore the potential for exploiting this contrast  
114 in the context of authentication. We begin in Study 1 by comparing authentication  
115 rates for legitimate account holders with authentication rates for i) zero-  
116 acquaintance attackers, and ii) personal attackers who know the participants very  
117 well (e.g. spouses, family members). In Study 2 we examine whether a full-  
118 visibility shoulder-surfing attack can be thwarted by presenting different  
119 photographs of the same targets to the participant and the attacker.

## 120 Study 1

121 The main aim of the first study was to establish whether participants could in  
122 practice generate suitable target faces. These should be faces that the  
123 participants know well, so that they could easily recognise them from  
124 photographs, but that other people do not know well, so that *all* of the faces in the  
125 array are unfamiliar to potential attackers. If such targets can be found, then it  
126 should be possible to differentiate between account holders and attackers by

127 comparing target detection performance. To anticipate, we found that suitable  
128 targets were readily volunteered by participants. Authentication rates were very  
129 high for legitimate users, even after a delay of one year. In contrast,  
130 authentication rates for attackers were very low, even when the attackers were  
131 close acquaintances of the users.

## 132 **Method**

### 133 *Participants*

134 A total of 396 volunteers contributed data. 120 were volunteers who responded to  
135 our recruitment email (54 male, 66 female; age range 18-79). These 120  
136 volunteers served as account holders in the current study. A further 110  
137 volunteers were recruited from our participant pool to act as zero-acquaintance  
138 attackers, that is, people who knew nothing about the account holders. For  
139 comparison, we also asked each account holder to nominate two close  
140 acquaintances (e.g. spouses, family members) who could act as personal  
141 attackers. We reasoned that the faces of people who are familiar to participants  
142 might also be familiar to their close acquaintances, giving these personal  
143 attackers a significant advantage. We acknowledge that this personal attacker  
144 selection is unrealistic, as it assumes that attacks only ever come from close  
145 acquaintances, and never from strangers. However, we prefer here to  
146 underestimate the security of the system than to overestimate it. 166 nominated  
147 attackers took part. All account holders and nominated attackers were offered  
148 entry into a prize draw for an iPod Nano. The study received ethical approval from  
149 the FIMS Faculty Ethics Committee at the University of Glasgow (CSE 00871).

### 150 *Design and Procedure*

151 The study consisted of seven distinct phases - three preparation phases and four  
152 test phases. We describe each of these below.

## 153 Phase 1: Target Nomination

154 Account holders nominated four or more target faces by entering the targets'  
155 names on the project website. As the proposed system relies on account holders  
156 and attackers having contrasting degrees of familiarity with the targets,  
157 appropriate selection of targets was critical. Ideally, an account holder's targets  
158 should be well known to the account holder, but unknown to other people. Our  
159 pilot work indicated that it can be difficult spontaneously to generate targets that  
160 satisfy both of these requirements. For this reason, we provided account holders  
161 with the following instructions in order to guide them to the appropriate region of  
162 their search space. Figure 3 represents the constraints on target selection  
163 schematically.

164 "The next page will ask you to list some minor celebrities - *really* minor  
165 celebrities.

166 Almost everyone recognises the 'A-List' celebrities below [photos of  
167 international celebrities such as major film stars]. Most people also recognise  
168 some 'B-List' celebrities [photos of national celebrities such as television  
169 presenters]. We want you to tell us your 'Z-List' celebrities.

170 By 'Z-List' celebrities, we mean people who are (or were):

- 171 1. Only famous within a narrow field of interest. For example, a famous skier  
172 or a famous cellist. This could include someone who was famous many years  
173 ago, but who is not well known these days.
- 174 2. Well known to you, so that you would easily recognise them from  
175 photographs.
- 176 3 Not well known to the public at large, so that you would not expect others to  
177 recognise them.
- 178 4. Possible to find using a Google Image search.

179 This is the most challenging part of the study, but also the most important."

180 Having read these instructions, account holders were asked to submit the names of  
181 four or more targets, up to a maximum of ten. There was no time limit for this task,  
182 and account holders were free to log out and return later to complete it. Once an  
183 account holder was satisfied with this personal list, the names were transferred to  
184 the experimenter. Each account holder was also asked to provide email addresses of  
185 two close acquaintances (e.g. spouses, family members) who would be willing to act  
186 as personal attackers.

### 187 Phase 2: Image Collection

188 Targets who had already been nominated by another account holder (<1%) were  
189 eliminated to avoid ambiguity at login. For all other targets, the experimenter  
190 collected at least four face photographs by using the target's name as a Google  
191 Image search term. We accepted the first four photographs in which the whole face  
192 was visible, regardless of viewing angle, lighting, age, or other sources of image  
193 variability. This resulted in 4 different photographs for each of 603 faces (2412  
194 images in total). All photos were cropped to a rectangular frame measuring 100  
195 pixels wide x 119 pixels high for presentation. The collected photos of each account  
196 holder's targets were then uploaded to the project website for that account holder to  
197 approve.

### 198 Phase 3: Image Approval

199 Account holders returned to the website to view the photos of their targets and to  
200 approve or decline each image. The purpose of this step was twofold. First, it  
201 allowed us to ensure that the photos depicted the correct individual. This was  
202 necessary as names are rarely unique identifiers, and search results invariably  
203 included images of more than one person. Second, it allowed us to confirm that the  
204 returned images were indeed recognisable to their nominators. Declined images  
205 (<1%) were replaced until the account holder was satisfied with the selection. Image  
206 approval was followed by a delay of one week to allow forgetting of the selection

207 procedure. Account holders then received an email requesting them to return to  
208 identify their faces again.

209 Phase 4: Account Holder Login (one week delay)

210 After the one-week delay, account holders returned to the project website and  
211 attempted to authenticate. The account holder's lock consisted of a series of four  
212 different challenge sets, each comprising nine face photographs arranged in a 3x3  
213 grid (similar to Passfaces challenge sets; see Figure 4). In each grid, one image (the  
214 target) was a random photo of a person selected at random from that account  
215 holder's pool of target names. The remaining eight images (the distractors) were  
216 random photos of faces drawn at random from other account holders' pools of  
217 targets. Allocation of the nine images to the nine grid positions was randomised so  
218 that location was not predictive of target/distractor status. This meant that from the  
219 perspective of the account holder, each grid contained one familiar face among  
220 eight unfamiliar faces. However, from the perspective of an attacker, all nine faces  
221 should be unfamiliar. The account holder's task was simply to click on the familiar  
222 face in each grid. Identifying the correct image in all four grids resulted in successful  
223 authentication. The probability of opening the lock by chance alone was thus 1 in  
224 6561, or 0.015%, for this particular instantiation.

225 No feedback was given until the end of the four-grid lock, after which the account  
226 holder was told whether or not the authentication attempt was successful. If the  
227 attempt was unsuccessful, the lock was reset using newly selected photos, and the  
228 account holder was asked to try again. Following successful authentication, or three  
229 unsuccessful attempts, the account holder proceeded to a brief questionnaire  
230 concerning account holders' impressions of the system.

231 Phase 5: Zero-Acquaintance Attacker Entry

232 In small-scale pilot studies, we found that medium-acquaintance attackers (work  
233 colleagues) were never successful. To estimate the success rate in a larger sample,  
234 we recruited 114 zero-acquaintance volunteers to attack a randomly allocated lock.

235 These 114 volunteers undertook 207 attacks between them. The authentication  
236 procedure for the attacker phase was exactly the same as for the account holder  
237 phase, with one of the account holder's targets and eight non-target faces making  
238 up each grid. As with the account holder entry phase, no performance feedback was  
239 given until successful authentication, or three unsuccessful attempts. We expected  
240 that if the account holder chose appropriate targets, none of these faces should be  
241 familiar to the attacker, and the success rate should not exceed chance levels. The  
242 zero-acquaintance attackers were recruited to verify that this was the case.  
243 However, our main interest was in the success rate of the personal attackers.

#### 244 Phase 6: Personal Attacker Entry

245 In the first phase of the study, each account holder was asked to provide email  
246 addresses of two close acquaintances who would be willing to act as personal  
247 attackers. A total of 166 personal attackers agreed to take part, undertaking 249  
248 attacks between them. Importantly, attackers only attacked their own nominator, so  
249 that every attack was from a close personal acquaintance of the account holder (e.g.  
250 spouse, family member), rather than from a stranger. Again, the authentication  
251 procedure was the same as for the account holder phase. If the account holder  
252 chose appropriate targets, all of the faces in all of the grids should be unknown to  
253 the attacker. We reasoned that high-acquaintance attackers might have acquired a  
254 degree of familiarity with their nominators' targets, due to shared exposure (e.g.  
255 overlapping interests or media consumption), thus providing a more stringent test.  
256 As with the account holder entry phase, no performance feedback was given until  
257 the end of the entire four-grid sequence that comprised a single lock. Following  
258 successful authentication, or three unsuccessful attempts, the attacker proceeded to  
259 a brief questionnaire concerning attackers' impressions of the system.

#### 260 Phase 7: Account Holder Login (One year delay)

261 One year after the initial account holder login phase, account holders were asked to  
262 authenticate a second time. This was the only contact between experimenters and

263 account holders since the initial login phase, and our log confirmed that none of the  
264 participants had visited the project website during the intervening months. Thus, the  
265 one year interval provided an excellent opportunity for account holders to forget  
266 about the study (Ebbinghaus, 1964). Previous research has shown that passwords  
267 are quickly forgotten once they fall into disuse (Witty & Brittain, 2004). For example,  
268 two studies of password memorability (Zviran & Haga, 1990, 1993) reported  
269 memorability rates of 35% and 27.2% after a delay of five months. Given that a  
270 putative advantage of our familiarity-based approach is that it imposes no memory  
271 load, we predicted relatively preserved authentication rates even after a year of  
272 disuse.

## 273 **Results and Discussion**

### 274 *Authentication data*

275 As can be seen in Table 1, 97.5% of account holders (117/120) successfully  
276 authenticated, with 84.2% (101/120) succeeding on the first attempt. In contrast,  
277 only 6.6% of personal attackers (11/166) were successful, and only 3.0% (5/166)  
278 on the first attempt. This compares favourably with previous analyses based on  
279 Passfaces (Davis, Monroe, & Reiter, 2004).

280 Chi Square analysis of these total success rates confirmed a highly significant  
281 difference between account holders and personal attackers [ $\chi^2(1) = 232.6, p < .$   
282 0001]. We also note that the majority of account holders' failures to authenticate  
283 were 'near misses', in which three of the four targets were correctly identified. For  
284 personal attackers, near misses were the least frequent authentication failure.

285 Only one attack by a zero-acquaintance attacker was successful, precluding any  
286 statistical analysis for this group. However, the circumstances of the one  
287 successful attack are perhaps revealing. Specifically, the account holder had not  
288 chosen 'Z-List' celebrities as required. Indeed, for the successfully attacked lock,

289 two of the four faces were members of the rock band Led Zeppelin (Robert Plant,  
290 Jimmy Page), perhaps analogous to choosing "ledzeppelin" as a password.

291 Analysis of the 11 successful attacks by nominated attackers revealed similar  
292 regularities. In five of these cases, the account holders had chosen widely-known  
293 celebrities as targets (e.g. Tony Blair, John Wayne), instead of 'Z-List' celebrities.  
294 In a further three cases, the account holders were non-Caucasian, and chose only  
295 non-caucasian target faces. Since virtually all of the distractor faces were  
296 Caucasian, these account holders' targets were presumably easy for their  
297 nominated attackers to guess. Nominated attackers were always close  
298 acquaintances of the account holders in this study, and so knew the ethnic  
299 background of the account holders they were attacking. For the remaining three  
300 successful attacks, we suggest that the attackers had some degree of familiarity  
301 with their account holders nominated targets - enough to set the targets apart  
302 from the distractors. For example, musicians that one likes might be recognized  
303 by one's spouse, due to shared exposure.

304 Taken together, the success rates of account holders (97.5%), randomly zero-  
305 acquaintance attackers (<1%), and nominated high-acquaintance attackers  
306 (6.6%) strike us as a promising starting point. Analysis of successful attacks  
307 provides little evidence that the principle of exploiting familiarity contrast is  
308 problematic. Rather, the main challenge is the separable problem of compliance:  
309 If the system is not used as intended, it does not work as well. This limitation is  
310 characteristic of a wide range of security systems - including passwords, PIN  
311 codes, and mechanical locks.

312 *Delayed Authentication*

313 79 of our initial account holders returned to login a second time, following a one-  
314 year delay. As can be seen in Table 1, 86.1% of these returning account holders  
315 (68/79) successfully authenticated, 78.5% (62/79) on their first attempt. This is a  
316 remarkably well-preserved success rate over such a long period of disuse, especially  
317 given that different images of the account holders' targets were presented at the  
318 delayed login. For comparison, previous research reported a first-attempt  
319 authentication rate of 77% after only two weeks when using traditional passwords  
320 (Bunnell et al., 1997). Established graphical authentication systems are also  
321 vulnerable to memory decay, though generally to a lesser degree than passwords.  
322 One influential study (Valentine, 1998) reported an authentication rate of 72% (by  
323 third attempt) after a five-month delay when using Passfaces. Although these  
324 comparisons involve rather different authentication methods, they highlight the very  
325 different demands of recall-based, recognition-based, and familiarity-based  
326 decisions.

327 We attribute account holders' high success rate in the present study to two main  
328 factors. First, there was no authentication code to remember, so the classic problem  
329 of account holders forgetting authentication codes did not apply. Second, our  
330 account holders had already established robust mental representations of their  
331 target faces prior to the study (they were familiar faces), so presenting different  
332 images of these targets did little to impede recognition (Jenkins & Burton, 2011).  
333 Interestingly, a number of returning account holders commented on the surprising  
334 ease of authentication under these conditions. One wrote, "I didn't think I could log  
335 in because I couldn't remember any of the people I chose - but I did!" Interestingly,  
336 another reported, "I got them all right. Did you use the same images of the people or  
337 different ones? I got the impression that I did not recognise the image but the  
338 person."

339 *Account Holders' Questionnaire Data*

340 Account holders responded to five questionnaire items concerning user  
341 experience. Summaries of these responses can be seen in Table 2.

342 The questionnaire data contain little evidence that account holders had difficulty  
343 using this system. None of the account holders reported writing down their  
344 targets' names. This suggests that they correctly understood that forgetting their  
345 targets was not an issue. Only 10% of account holders reported difficulty in  
346 identifying their target faces. Thus most account holders were successful in  
347 nominating faces that they could recognise well. Interestingly, the great majority  
348 of account holders (80%) stated that with the benefit of hindsight, they would  
349 have chosen different targets. Presumably, since account holders had little trouble  
350 recognising targets that they actually chose, their motive here was not making  
351 authentication easier for themselves, but making it harder for attackers. 16% of  
352 account holders reported recognising one of the non-target faces in a grid.  
353 However, the overall authentication rate of 97.5% implies that this confusion  
354 rarely stopped them from authenticating correctly. On the basis of this  
355 experimental trial, 31% of participants said that they would use a Facelock system  
356 instead of a password, 25% said they would not, and 44% were undecided. Given  
357 that we made no concessions to usability and HCI issues in this study, it is  
358 perhaps surprising that 31% of respondents were positively disposed to the  
359 method.

#### 360 *Personal attackers' Questionnaire Data*

361 Personal attackers responded to four questionnaire items using a 5-point Likert  
362 scale, where 1 indicates a low rating, and 5 indicates a high rating. Mean ratings  
363 for each item are shown in Table 3.

364 Personal attackers found guessing their account holders' targets moderately  
365 effortful, and found it quite difficult to imagine who the account holder might have  
366 chosen. Consistent with these impressions, they rated their level of success as  
367 rather poor overall, though even this rating is a generous appraisal of their actual

368 success rate. Personal attackers knew their account holders very well overall,  
369 confirming good compliance among account holders at the attacker nomination  
370 stage. To test whether personal attackers were more successful the better they  
371 knew their victims, we computed the correlation between these attackers'  
372 acquaintance ratings for Item 4 above, and the number of correctly-guessed  
373 targets (0-4) in their first attacks (see Figure 5).

374 This correlation was moderately positive and highly reliable [ $R = 0.29$ ,  $N = 166$ ,  $p$   
375  $< 0.001$ ]. Importantly, lower acquaintance attackers (ratings  $< 4$ ) were never  
376 successful. We return to the issue of acquaintance in the General Discussion  
377 section.

## 378 **Study 2**

379 The preceding study confirmed that account holders who were familiar with the  
380 target faces could easily distinguish these faces from unfamiliar non-targets,  
381 regardless of the particular photos that were used to portray them. In contrast,  
382 attackers found it very difficult to guess account holders' targets, even when the  
383 attackers were close acquaintances of the account holders.

384 The second study focuses on a different aspect of the proposal, specifically the use  
385 of multiple photos of each target. We also sought to compare the resilience of  
386 different account holders' locks directly, by exposing them to multiple attacks. To  
387 this end, we modelled a best-case scenario for shoulder-surfing attacks, in which we  
388 presented the correct authentication sequence to attackers under full-visibility  
389 viewing conditions, and then asked them immediately to replicate the sequence  
390 using different photographs of the same target faces. Attackers were thus required  
391 to generate the sequence of identities that they had just seen, even though those  
392 identities were portrayed using different images.

393 As in Study 1, we loaded this situation heavily in the attackers' favour. First, we used  
394 the same four target identities for the observation sequence and the replication  
395 sequence, rather than drawing a set of four targets at random from the account

396 holders' entire pool. Second, we presented these same four targets in the same  
397 order in both sequences, rather than presenting them in a different random order  
398 each time. Third, attackers did not have to glance furtively at the authentication  
399 sequences for fear of being noticed. Instead, we presented the sequences very  
400 clearly to the attackers, who were asked to give it their full attention. Finally, there  
401 was no delay between the observation sequence and the replication task. Thus  
402 attackers' memory decay was minimized. These real world complications were  
403 eliminated in an effort to isolate the impact of a photo change. It is already  
404 established that replicating a four-item sequence is well within the limits of human  
405 short-term memory. This is true in experimental settings (Miller, 1956), and also in  
406 the context of shoulder-surfing 4-digit PIN numbers (Anderson, 1993). However, the  
407 present case differs from previous studies in that different images of each item are  
408 used at the sequence replication stage. If attackers are able to integrate across  
409 different photos of each target efficiently, then performance should be close to  
410 ceiling (Miller, 1956). Alternatively, if a change in photograph impedes identification  
411 in this situation, then performance should be relatively poor, even when the  
412 authentication code is clearly presented to the attacker immediately before the  
413 attack.

#### 414 **Method**

##### 415 *Participants*

416 Thirty-two postgraduate volunteers (6 male, 26 female; age range 21-36) completed  
417 the study. The study received ethical approval from the FIMS Ethics Committee at  
418 the University of Glasgow.

##### 419 *Design and Procedure*

420 Each participant attacked five locks so that each lock was attacked 32 times. The  
421 five locks (i.e. 5 different 4-grid sequences) were drawn at random from those that  
422 led to successful authentication by account holders in Study 1. In other words, the

423 authentication rate for account holders was 100% for this sample of locks. For each  
424 lock, a different-image version was also constructed, by replacing the target from  
425 each grid with a different photo of the same person, and replacing the eight non-  
426 targets with different non-targets.

427 As with the original grids, the location of the images in the grid was randomised. To  
428 make the task as easy as possible for the attackers, grid order was preserved across  
429 observation and replication sequences, so that the same targets appeared in the  
430 same order (1-4) in both versions of the lock. The different-image versions of the  
431 grids were printed at a size of 10 cm x 12 cm and bound into response booklets. The  
432 original grids were projected at a size of 150 cm x 180 cm using a computer  
433 controlled data projector, which attackers viewed at a distance of between 3 and 5  
434 metres.

435 For each of the five locks, attackers first watched the authentication sequence using  
436 the original grids, and then tried to replicate the sequence on the different-image  
437 grids, that is, to copy the account holder's authentication code. To demonstrate each  
438 sequence as clearly as possible, each one of the four grids was presented on screen  
439 for 5 seconds together with its grid number (1-4). After the first 2 seconds, a green  
440 frame appeared around one of the faces, identifying that face as the target  
441 (analogous to watching the account holder select that face). As face identification is  
442 normally accomplished within about 200 msec of stimulus onset (Liu, Harris, &  
443 Kanwisher, 2002), we expected this presentation time to allow full encoding of the  
444 correct target. This procedure was intended to model observation of target selection  
445 in an optimal shoulder surfing situation, in which all the necessary information is  
446 presented clearly at the focus of attention. Readers are invited to simulate this task  
447 for a single grid by comparing Figure 6 and Figure 4.

448 Successive grids in each lock were separated by a blank interval of 2 seconds.  
449 Immediately after the fourth target had been revealed, attackers were asked to  
450 reproduce the sequence they had just seen, by circling the same four targets on

451 their response sheets. There was no time limit for this task. When the attackers were  
452 ready to proceed (<60 secs in all cases), the next authentication sequence was  
453 initiated. All 32 participants attacked the same 5 locks once, resulting in 160 attacks  
454 in total.

## 455 **Results and Discussion**

456 Raw frequency data are shown in Table 4. Only 3 out of 160 attacks were successful  
457 (1.9%). This strikes us as a very promising figure, especially given the privileged  
458 conditions of attack. When attempting to replicate the authentication sequence,  
459 attackers saw the same targets presented in the same order under highly favourable  
460 viewing conditions and with no time pressure. Only the photo used for each face was  
461 changed. As it turned out, this alone was enough to defeat these shoulder-surfing  
462 attacks.

463 We note that all three successful attacks were on the same lock. Inspection of the  
464 targets in this particular lock suggests that this may be due to their distinctive  
465 appearance. For example, one of the targets was bald and wore glasses in both  
466 photos; another was an elderly woman with permed white hair. As none of the  
467 distractor faces shared these features, the matching targets were presumably rather  
468 salient in this context. In the General Discussion we consider how this situation  
469 could be avoided.

## 470 **General Discussion**

### 471 *Summary of Findings*

472 Two studies tested a knowledge-based authentication system that exploits the  
473 psychological contrast between familiar and unfamiliar face recognition. In Study 1  
474 we found that account holders were able to generate target faces that were well  
475 known to themselves, but were not well known to other people. Account holders  
476 authenticated easily by detecting these familiar targets among other faces (97.5%  
477 success rate), and this was the case even after a one-year delay (86.1% success

478 rate). By contrast, zero-acquaintance attackers were reduced to guessing (<1%  
479 attacks rate). Even personal attackers who knew the account holder well were rarely  
480 able to authenticate (6.6% success rate). This success rate for attacks compares  
481 favourably with previous studies. Analysing a system based on Passfaces, Davis et al  
482 (2004) conclude that 10% of authentication codes could be guessed within one or  
483 two attempts, even by very low acquaintance attackers who know only the gender  
484 or race of the account holder. Here we found a successful attack rate of 6.6% within  
485 three attempts for very high acquaintance attackers who knew a great deal about  
486 the account holder. In Study 2 we found that optimal shoulder-surfing attacks by  
487 strangers could be repelled simply by using different photos of the targets in the  
488 observed and attacked grids (1.9% success rate). Together, these findings suggest  
489 that the contrast between familiar and unfamiliar face recognition may be useful for  
490 graphical authentication systems. Although face-based systems have been  
491 developed previously, these have always conflated face recognition and image  
492 recognition, by representing each face with a single image (Jenkins, White, van  
493 Montfort, & Burton, 2011). As image memory will be equally excellent for account  
494 holders and attackers, such systems are vulnerable to shoulder-surfing attacks (Tari,  
495 Ozok, & Holden, 2006). The use of different photographs for each target confounds  
496 attackers who are unfamiliar with the targets, but does not impede legitimate users  
497 who are familiar with their chosen targets.

498 The approach we describe here offers two advantages. First, unlike a conventional  
499 password, it does not require the account holder to remember anything specific to  
500 the authentication procedure, as the task is simply to indicate which of several faces  
501 is familiar. The system thus exerts very little memory load compared with  
502 conventional passwords. Our most striking evidence for this comes from the delayed  
503 authentication task in Study 1. Here, account holder's authentication rate was 86%,  
504 one year after a single login. This is unprecedented for knowledge based  
505 authentication systems (Sasse, Brostoff, & Weinrich, 2001). For comparison, one  
506 evaluation of traditional passwords reported authentication rates of 27% after just 3

507 months (Zviran & Haga, 1993). A similar evaluation of Passfaces found  
508 authentication rates of 72% after 5 months (Valentine, 1998). Such studies  
509 contribute to the general finding that memory decay impacts image recognition less  
510 than it impacts password recall. Here we show that memory decay impacts face  
511 familiarity judgements even less. Second, it does not matter greatly if authentication  
512 is observed. As Study 2 shows, even when an attacker sees the same set of targets  
513 when attempting to authenticate, authentication is still difficult when different  
514 photos of those targets are presented. Previous work has shown that Passfaces is  
515 highly vulnerable to shoulder surfing when a mouse pointer is used to select targets.  
516 Participants in that study rated the vulnerability of Passfaces at 5.2 on a scale from 1  
517 (not vulnerable) to 7 (extremely vulnerable), indicating that shoulder surfers found it  
518 very easy to obtain the faces by observation. In the same study, dictionary based  
519 passwords were rated 4.85 in terms of vulnerability. Interestingly, using a keyboard  
520 instead of a mouse to select targets reduced the vulnerability of Passfaces from 5.2  
521 to 2.3, presumably because keyboard entry forced onlookers to divide their attention  
522 between the screen and the keyboard (Braun, 1998). For the same reason, keyboard  
523 input should strengthen the scheme we propose here.

#### 524 *Limitations*

525 Our testing exposed a number of important limitations to the system in its  
526 experimental form. First, the lock is vulnerable to an attacker who, like the account  
527 holder, knows the target faces. This was evident in Study 1, in which attackers who  
528 were closest acquaintances of the account holders correctly guessed more targets  
529 than attackers who were less close acquaintances. This vulnerability underscores  
530 the importance of appropriate target selection. One way for a secret holder to  
531 minimise risk would be to maintain a large pool of target faces, and to sample these  
532 from disparate fields of interest, so that no single attacker knows enough targets to  
533 authenticate.

534 A second limitation is that attackers may be able to match different images of  
535 targets whose appearance is both distinctive (e.g. bald head and round glasses), and  
536 stable (i.e. similar appearance in all photos). This was seen in Study 2, where one  
537 lock that contained highly distinctive faces could be compromised in a shoulder-  
538 surfing attack. For similar reasons, target distinctiveness may be a concern  
539 whenever an account holder's targets are all drawn from a single ethnic group or  
540 age band. These risks could be reduced by avoiding highly distinctive faces, and by  
541 avoiding similar images of any particular target.

542 De Angeli et al. (2005) proposed that graphical authentication mechanisms such as  
543 Facelock should be assessed in terms of guessability, observability and recordability  
544 when considering how they can be breached. Table 5 shows a threat model based on  
545 this taxonomy.

#### 546 *Future Directions*

547 One pragmatic concern is scalability. Our experimental implementation of Facelock  
548 involved a multi-step enrollment process, and required considerable human labour  
549 to find images of targets and verify these with the account holders. This may not be  
550 feasible for a large-scale system. Unless these steps can be significantly  
551 streamlined, the approach may be better suited to small-scale or personal  
552 deployments such as locking computers and mobile devices than to large-scale  
553 deployments such as securing bank accounts.

554 The studies we report here suggest a number of possible directions for future  
555 development. One would be to select non-targets automatically for each grid based  
556 on their similarity to the target. For example, if the target for a particular grid is a  
557 young Asian female, the non-targets used to complete that grid could also be young  
558 Asian females. Increasing the homogeneity of the grids should undermine attacks  
559 that rely on distinctiveness to infer targets (Study 1). This functionality would require

560 all images in the system to be tagged with properties such as age, sex, and race.  
561 Automatic tagging is currently a major focus of image analysis (Datta, Joshi, Li, &  
562 Wong, 2008), and much progress has been made in recent years (see Bengio, 2009,  
563 for an instructive overview). Indeed, human similarity ratings of faces can already be  
564 accurately predicted by automatic systems (Lacroix, Postma, & Murre, 2005), which  
565 could dramatically improve the effectiveness of facelock image arrays.

566 We noted in Study 1 that 80% of account holders would choose different targets if  
567 they could choose again. As authentication failures were so rare among these  
568 account holders, it seems reasonable to assume that they would not have chosen  
569 different targets to make their own authentication even easier, but rather to make  
570 fraudulent access even harder. Presumably faces that are less widely known  
571 occurred to these account holders after the study had begun, and the account  
572 holders realised that these would make better targets. If so, allowing account  
573 holders to update their pool of target faces could improve the security of the system.

574 A related issue concerns the optimal number and set size of the grids that are used  
575 to authenticate. In the present studies we arbitrarily chose a sequence of four 3 x 3  
576 grids, which corresponds to a guessing rate of 1 in 6,561. It would be technically  
577 trivial to change the guessing rate by changing the grid configuration (e.g. 1 in  
578 1,048,576 for 5 different 4 x 4 grids), but implementation details are not our priority  
579 here. Our main concern is whether familiarity contrasts in face recognition may be  
580 exploited to improve the security of authentication systems. This question is  
581 independent of any particular grid configuration. Dedicated usability studies will be  
582 required to examine trade offs between security and ease of use. Such studies  
583 should also seek to optimise task instructions to make them as easy as possible to  
584 follow. In Study 1, five of the eleven successful attacks from personal attackers, and  
585 the single successful attack from a random attacker, were all attributable to account  
586 holders nominating major celebrities as their targets, despite instructions to the  
587 contrary. Clearer instructions, or tighter constraints on the target nomination  
588 process, could mitigate this vulnerability.

589 *Concluding remarks*

590 Although we have outlined a novel approach to graphical authentication using faces,  
591 there are clearly very many issues outstanding. In this final section we highlight  
592 some of these in the hope that we can be as clear as possible in articulating what is  
593 and is not claimed for this proposal.

- 594 • We are not presenting Facelock as a packaged product that is ready to deploy.  
595 Instead we offer these initial studies as proof of principle. Our focus  
596 throughout is on the familiarity of a face to the observer, and how this  
597 profoundly affects the observer's ability to process images of that face. The  
598 key contrast between familiar and unfamiliar face perception has seldom  
599 been addressed in the computer science literature (Sinha, Balas, Ostrovsky, &  
600 Russell, 2006). Here we hope to have demonstrated that this contrast may be  
601 usefully exploited in graphical authentication systems. However, a number of  
602 usability issues (discussed above) would need to be resolved before such a  
603 system could be practically deployed.
- 604 • We do not claim that the proposed system is flawless. In the studies we  
605 present, some account holders failed to authenticate, and some attackers  
606 succeeded. We address both of these outcomes, alongside other limitations of  
607 the studies, in the discussion section of the paper. Our main emphasis is the  
608 relative performance of observers who are familiar or unfamiliar with the  
609 faces concerned. In perceptual experiments, recognition performance is  
610 radically different for these two groups. Here we show that the same applies  
611 when the task is incorporated in an authentication system.
- 612 • We are not claiming that Facelock is superior to Passfaces. Any such  
613 evaluation would require a direct comparison of the two approaches, and we  
614 have not attempted that here. Previous studies have looked at memorability  
615 of Passfaces (Valentine, 1998) and its susceptibility to shoulder-surfing attacks  
616 (Tari, Ozok, & Holden, 2006), and we consider these issues also. However,

617 Passfaces is an established commercial system. Facelock, as an experimental  
618 proposal, is unfettered by implementation concerns. Any attempt to compare  
619 performance directly would thus be rather unfair on Passfaces. Indeed, the  
620 general question of which system authentication system is 'best' is likely too  
621 simplistic. Any approach will have its own profile of strengths of weaknesses,  
622 and will be better suited to some situations - and to some users - than to  
623 others.

- 624 • We do claim that it is easy for users to generate a set of faces that are well  
625 known to them, but not to other people. We show that an authentication code  
626 based on such faces makes it easy for the user to login, even after a year of  
627 disuse, as it does not require the user to commit anything to memory. The  
628 user's authentication code is difficult for other people to guess, even for close  
629 acquaintances such as spouses. It is also highly resistant to shoulder-surfing,  
630 as image changes that are transparent for the (familiar) user are not  
631 transparent for the (unfamiliar) attacker.

632 More generally, we propose that research into graphical authentication systems can  
633 exploit findings from psychological research, and that psychological research can be  
634 enriched by considering applied problems in other fields. Image recognition is not  
635 the same as face recognition. Unfamiliar face recognition is not the same as familiar  
636 face recognition. Not all observers are equal. These insights offer much scope for  
637 innovation in face-based graphical authentication systems, and we hope that the  
638 current studies might spur further development in this direction. **References**

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# Figure 1

Matching a face to a poor quality CCTV image

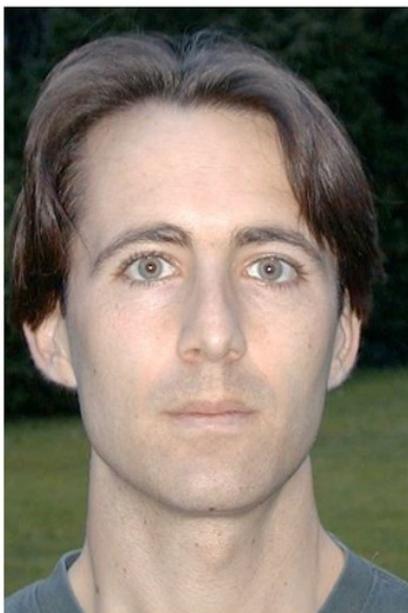
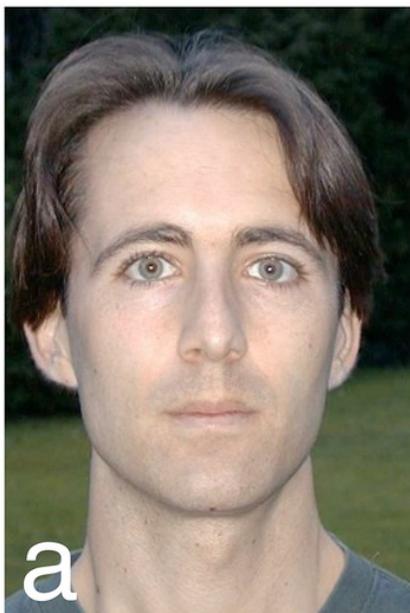
Figure 1. Example images from Burton, Wilson, Cowan, & Bruce (1999). Matching poor quality images is easy for observers who are familiar with the faces concerned. Performance of unfamiliar observers is strikingly poor. These images both show the same person.



# Figure 2

Familiar and unfamiliar face matching

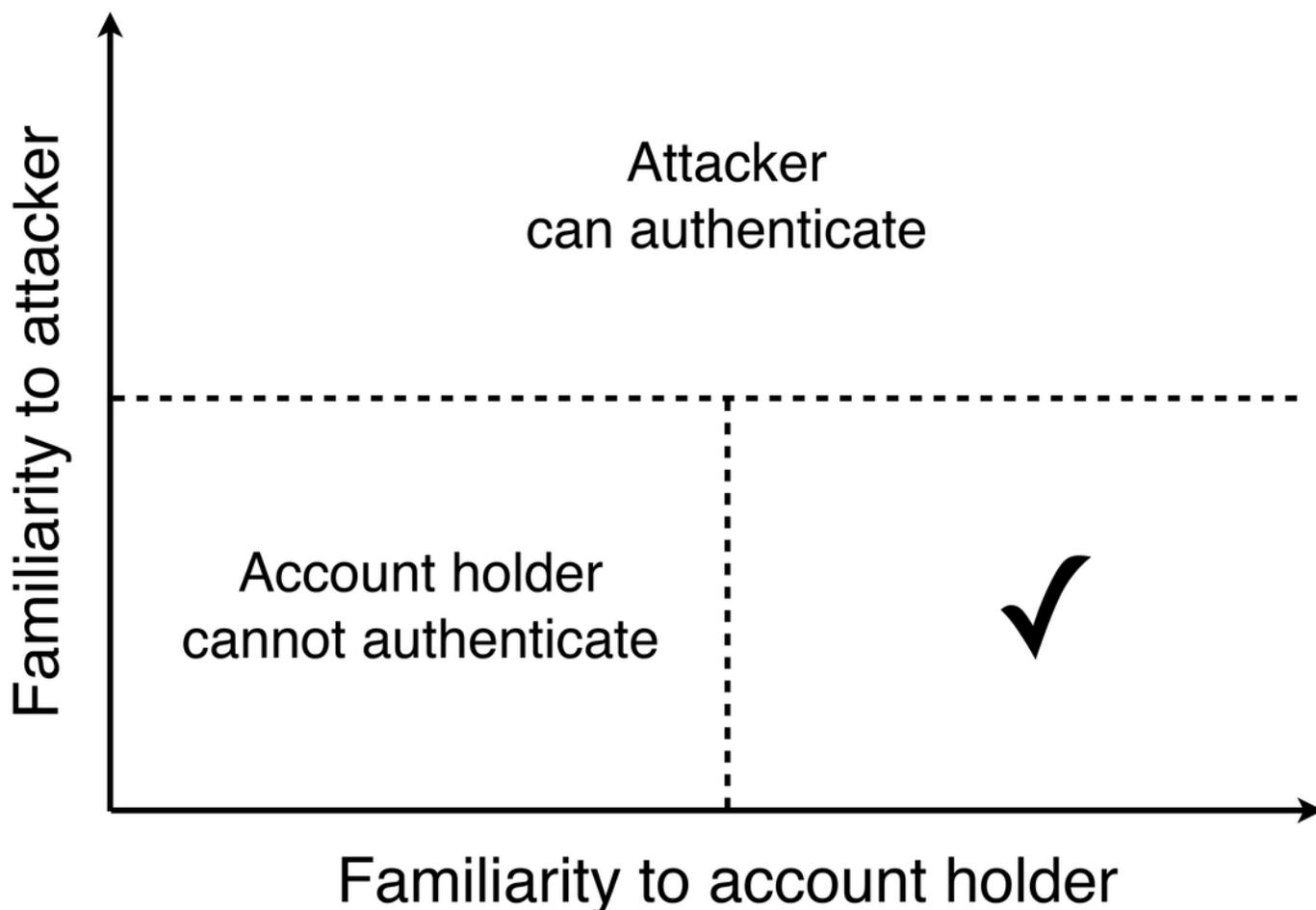
Figure 2. (a) Matching identical images is trivial. (b) Matching different images of unfamiliar faces is hard. (c) Matching different images of familiar faces is easy.



# Figure 3

## Selecting Facelock targets

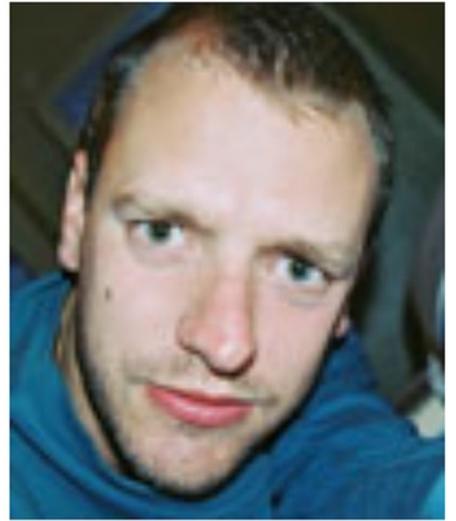
Figure 3. A schematic diagram summarising the requirements of target faces. If the target is familiar to the attacker, the attacker will be able to authenticate. If the target is unfamiliar to the account holder, the account holder will be unable to authenticate. The tick represents the region of acceptable targets.



# Figure 4

A challenge grid in Facelock

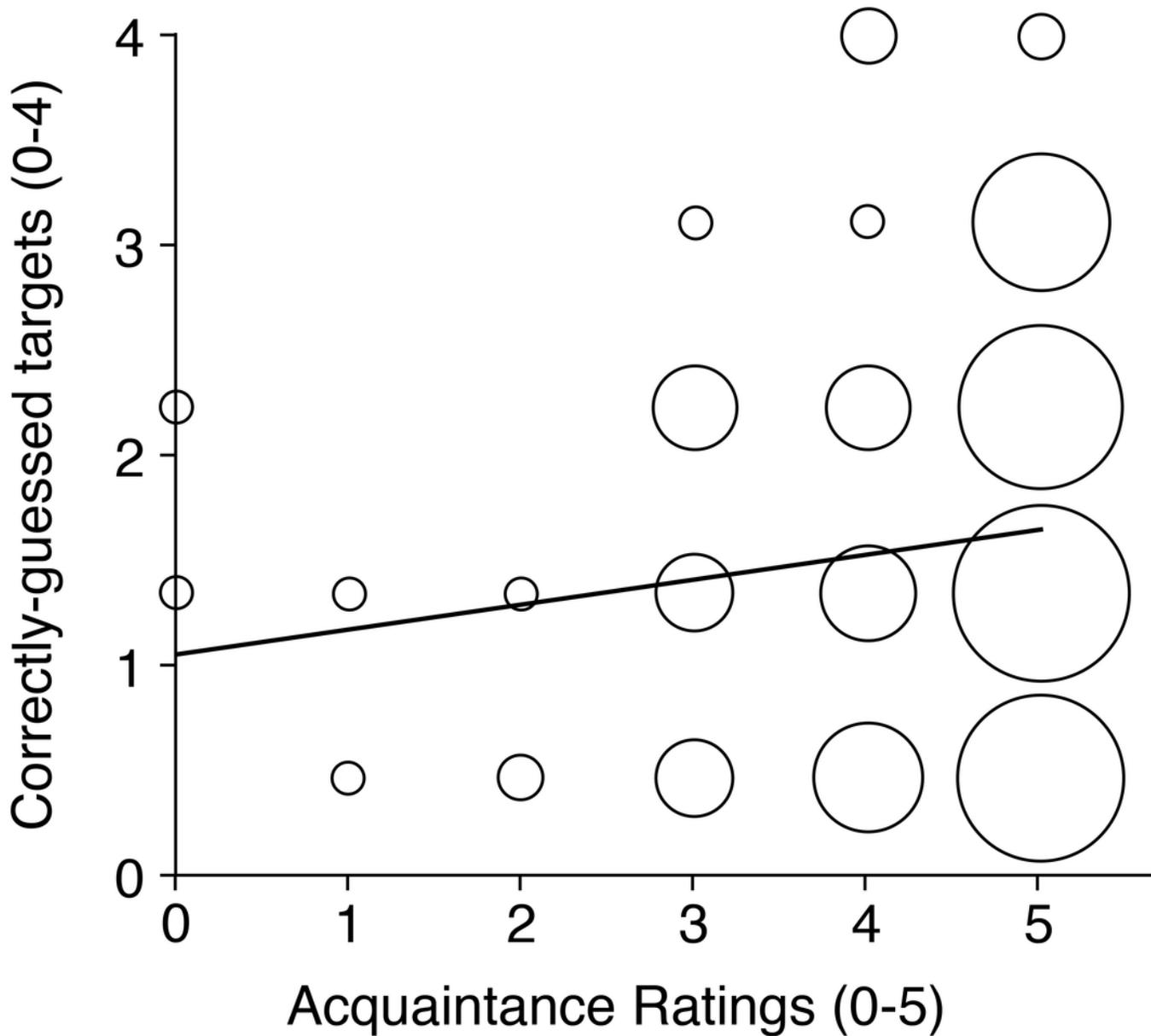
Figure 4. An example grid consisting of one face (the target) that is familiar to one of our account holders, and eight faces that are unfamiliar to the same account holder. Readers are invited to guess which of the nine faces is the target. For someone who doesn't know the account holder, it is difficult to find any basis for this decision.



# Figure 5

Attack success as a function of personal acquaintance

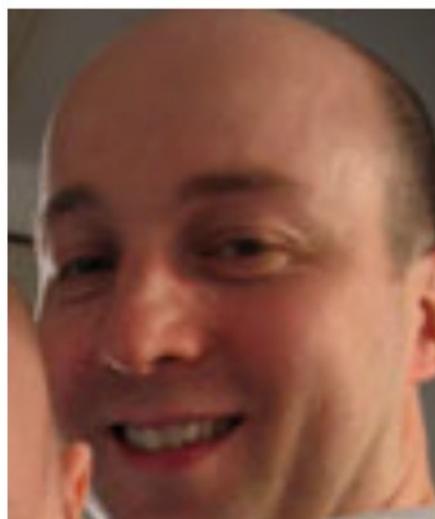
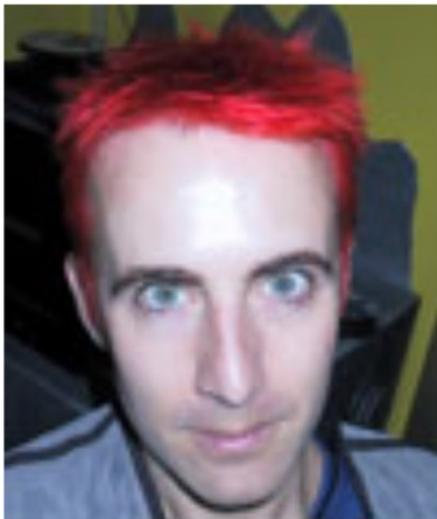
Figure 5. Scatterplot showing the relationship between Personal attackers' Acquaintance Ratings and the number of correctly guessed targets in their first attacks. The area of each datapoint is proportional to the number of cases contributing to it.



# Figure 6

A second challenge grid

Figure 6. One of these faces is also present in Figure 4. Even with a single grid, it is difficult to determine which face is repeated simply by trying to memorise Figure 4. Side-by-side matching of unfamiliar faces is also highly error prone (Jenkins & Burton, 2008, 2011).



**Table 1** (on next page)

Authentication rates in Study 1.

Table 1. Authentication success rates in Study 1, shown separately for Account Holders and Attackers. See main text for details of delays of procedure.

Table 1. Authentication success rates in Study 1, shown separately for Account Holders and Attackers. See main text for details of delays of procedure.

	<i>N</i>	Succeeded		Succeeded (1st attempt)		Failed	
Account holders (1 week delay)	120	117	97.5%	101	84.1%	3	2.5%
Account holders (1 year delay)	79	68	86.1%	62	78.5%	11	13.9%
Zero-acquaintance attackers	114	1	0.9%	0	0%	113	99.1%
Personal attackers	166	11	6.6%	5	3.0%	155	93.4%

**Table 2**(on next page)

Questionnaire data from Study 1.

Table 2. Percentage 'Yes' responses for Account Holders' Questionnaire items from Study 1.

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I wrote my targets' names down to remember them.	0%
I found it hard to identify my target faces.	10%
Upon reflection, I would have chosen different target faces.	80%
I was confused by recognising more than one face in a grid.	16%
I would be prepared to use a system like this to log in rather than a password.	31%

**Table 3**(on next page)

Attacker questionnaire data from Study 1.

Table 3. Mean Likert scale ratings (1-5) for Personal attackers' Questionnaire items from Study 1.

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How much effort was involved in guessing the targets?	2.9
How hard was it to put yourself into the account holder's shoes to guess his/her targets?	3.5
How successful do you think you were?	2.3
How well do you know the person?	4.4

**Table 4**(on next page)

Shoulder-surfing data from Study 2.

Table 4. Shoulder-surfing data from Study 2. Columns refer to the different locks, and rows refer to the number of correctly-guessed targets. All four targets must be correctly guessed for the attacker to gain entry.

Table 4. Shoulder-surfing data from Study 2. Columns refer to the different locks, and rows refer to the number of correctly-guessed targets. All four targets must be correctly guessed for the attacker to gain entry.

Correctly-guessed targets	Lock 1	Lock 2	Lock 3	Lock 4	Lock 5
0	9	0	5	15	9
1	10	4	14	10	13
2	11	16	12	7	9
3	2	9	1	0	1
4	0	3	0	0	0

## **Table 5**(on next page)

Threat model.

Table 5. A threat model for Facelock, based on De Angeli et al. (2005).

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Threat	Vulnerability	Attack Exploits	Facelock Mitigation
Guessability	Predictable choices	Knowledge of a user	Targets are minor celebrities
Observability	Ease of shoulder surfing	Observation of user selecting faces	Different images of different targets for each login
	Ease of intersection attacks	Refreshing the screen to see which face stays the same	Different images of different targets at each refresh Limited login attempts
Recordability	Ease of recording targets' names	User insecure behaviour	No incentive for account holders to write down target names
	Ease of recording the screen	Use of mobile phone cameras or screen shots	Different images of different targets for each login