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Sex differences in the use of social information emerge under conditions of risk

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Social learning provides an effective route to gaining up-to-date information, particularly when information is costly to obtain a-socially. Theoretical work predicts that the willingness to switch between using a-social and social sources of information will vary between individuals according to their risk tolerance. We tested the prediction that, where there are sex differences in risk tolerance, altering the variance of the payoffs of using a-social and social information differentially influences the probability of social information use by sex. In a computer-based task that involved building a virtual spaceship, men and women (N=88) were given the option of using either a-social or social sources of information to improve their performance. When the a-social option was risky (i.e., the participant's score could markedly increase or decrease) and the social option was safe (i.e., their score could slightly increase or remain the same), women, but not men, were more likely to use the social option than the a-social option. In all other conditions, both women and men preferentially used the a-social option to a similar degree. We therefore found both a sex difference in risk aversion and a sex difference in the preference for social information when relying on a-social information was risky, consistent with the hypothesis that levels of risk-aversion influence the use of social information.

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24 **Abstract**

25 Social learning provides an effective route to gaining up-to-date information, particularly when
26 information is costly to obtain a-socially. Theoretical work predicts that the willingness to switch
27 between using a-social and social sources of information will vary between individuals according
28 to their risk tolerance. We tested the prediction that, where there are sex differences in risk
29 tolerance, altering the variance of the payoffs of using a-social and social information
30 differentially influences the probability of social information use by sex. In a computer-based
31 task that involved building a virtual spaceship, men and women (N=88) were given the option of
32 using either a-social or social sources of information to improve their performance. When the
33 a-social option was risky (i.e., the participant's score could markedly increase or decrease) and
34 the social option was safe (i.e., their score could slightly increase or remain the same), women,
35 but not men, were more likely to use the social option than the a-social option. In all other
36 conditions, both women and men preferentially used the a-social option to a similar degree. We
37 therefore found both a sex difference in risk aversion and a sex difference in the preference for
38 social information when relying on a-social information was risky, consistent with the hypothesis
39 that levels of risk-aversion influence the use of social information.

40

41 Introduction

42 Individuals can acquire information either directly through their own asocial learning experiences
43 or by copying other individuals [1]. Asocial learning allows individuals to gain first-hand
44 knowledge about the immediate environment, but reliance on this type of learning can be costly,
45 for instance, in terms of time and energy [2]. In contrast, social learning can provide a cost-
46 effective route to gaining up-to-date information, particularly when the environment is changing
47 and information is costly to obtain asocially [2, 3]. Theoretical models support the hypothesis
48 that an increased reliance on social learning is adaptive when the environment becomes more
49 variable (although not when variability is very high) and when the returns from asocial learning
50 become more unreliable [3-6]. Therefore, individuals are predicted to be sensitive to the
51 reliability of the available sources of information and to use these reliability estimates when
52 choosing whether to learn asocially or socially [2].

53 Reliability can include the predictability of the source of information (e.g., the likelihood
54 that a food reward is associated with a particular cue) and the variability in the expected payoff
55 derived from different sources (e.g., the variability in the amounts of food obtained from different
56 foraging patches). Empirical research on non-human animals and humans has shown that
57 individuals are likely to use social learning when personal experience reveals that the
58 environment is unpredictable or the variability in payoffs of available options is high (e.g., [7-
59 10]). For example, a study of nine-spined sticklebacks (*Pungitius pungitius*) showed that
60 reducing the predictability of personally experienced cues in a foraging context increased
61 reliance on social learning [10]. Similarly, when faced with the option of taking a risky decision in
62 an experimental paradigm, human participants were found to delay their decision and observe
63 the choices made by others [7]. These findings support the broader hypothesis that social
64 learning is used strategically [11].

65 Individual differences in the use of social information are predicted to reflect individual
66 differences in risk tolerance [12]. Risk-averse individuals are expected to switch to using social
67 sources sooner than risk-prone individuals when faced with unreliable personal experience. In
68 real-world scenarios, the predictability and riskiness of sources of information are likely to co-
69 vary; for example, food items with high nutritional value are likely to be both rarer in the
70 environment, and more difficult to obtain, than low value food items [4, 13]. By switching to
71 social sources of information when faced with risky options, individuals are thus potentially
72 better able to exploit high-value resources. In both non-human animals and humans, individuals
73 vary in their sensitivity to experiencing gains and losses [14, 15], and a small number of studies
74 of non-human animals have revealed that 'shy' individuals are more likely than 'bold' individuals
75 to copy the decisions of others (e.g., [16-19]). However, the link between risk-proneness and
76 social information use has yet to be evaluated in humans.

77 One variable that is commonly related to risk aversion in humans is an individual's sex,
78 with women showing lower average scores than men on a range of measures of risky behaviour
79 (e.g., [20-22]). While the degree of overlap between the sexes on risk-aversion measures is
80 often considerable [23], and not all risk-aversion measures show sex differences [24], women
81 perceive the benefits gained from taking risks as being lower than do men [24, 25]. Women also
82 rate both the likelihood of a negative outcome and the perceived severity of the costs higher
83 than do men [24, 25], and report being less likely than men to engage in novel activities that
84 involve risk [22]. Similarly, data from personality measures indicate that, on average, women are
85 more sensitive than men to the potential negative outcomes of decisions [26]. The probability of
86 using asocial versus social sources of information when faced with a risky decision is therefore
87 likely to differ on average between women and men.

88 The aim of this study was to examine whether altering the riskiness of using asocial and
89 social sources of information would differentially influence the probability that men and women

90 used these sources. Here, we are defining riskiness in terms of variation in expected score [27].
91 We predicted that, when one of these sources of information appeared to be risky (i.e., high
92 variation in expected score), women would be more likely than men to use the alternative
93 source, safe (i.e., low variation in expected score) of information. In the control condition, no sex
94 difference in the use of asocial and social sources was predicted. We designed a novel
95 computer-based task that involved constructing a virtual spaceship. After building the first
96 spaceship, participants were given the option of using asocial or social sources of information to
97 improve their ranked score. Participants were assigned to one of three conditions, in which
98 either i) the asocial option was risky and the social option was safe, ii) the asocial option was
99 safe and the social option was risky, or iii) both the asocial and social options were safe.
100 Because scores were randomly allocated to spaceships, participants could not learn about the
101 usefulness of different design features. The outcome measure of principal interest was the
102 participant's choice of information source. Participants also completed a risky impulsivity
103 measure [28], in order to confirm that the predicted sex difference in average score was found in
104 our set of participants.

105

106 **Methods**

107 *Participants*

108 Eighty-eight participants (50 women and 38 men) were recruited through the School of
109 Psychology & Neuroscience's online participant recruitment system. All participants were aged
110 17 or over, with the majority (91%) falling into the 18-25 age range. Participants gave consent
111 before taking part in the experiment and were debriefed afterwards. All participants were
112 reimbursed £3 for attending the session – which lasted approximately 20 minutes – and could
113 obtain an additional £2 depending on performance criteria (see 'Procedure' below). Participants

114 were randomly assigned to one of the three conditions (see 'Asocial and social information') and
115 were tested in groups (range = 4-9 individuals). Participants gave consent via a button click at
116 the start of the experiment. This was approved, as were all other procedures used in this study,
117 by the Ethics Committee of the School of Psychology & Neuroscience on behalf of the
118 University of St Andrews (approval code PS11481).

119

120 *Procedure*

121 Participants stated their gender ('female', 'male', 'other' or 'prefer not to say') and age bracket
122 before beginning the experiment. They then played a computer game, programmed using web-
123 based JavaScript, in which they built virtual spaceships. Participants were instructed that the
124 aim was to construct spaceships with the highest scores, and that the participant with the
125 highest score at the end of the session would receive a bonus payment. Spaceship construction
126 proceeded in three rounds, each with two building phases. In Phase 1, participants constructed
127 their first spaceship by selecting tiles from a grid of thirty available items that were arranged into
128 themes (crew, cargo, engines, shields and lasers) (see **Figure 1a**). Players had two minutes to
129 view these items and choose ten to place on a spaceship template. The only constraint was that
130 they had to use at least one crew member and one engine. After finishing Phase 1, each
131 player's ship was given a numerical score and a rank in a league table (1st to 5th, highest to
132 lowest) (see **Figure 1b**). Players were given no information on how a good score might be
133 achieved, and, in reality, scores were randomly assigned to the participants' spaceship (with a
134 range of 8,000 to 25,000), along with a false 'rank' that was always either 1st, 3rd or 5th.

135 Participants then chose between using asocial and social sources of information (see
136 'Asocial and social options') before building a second spaceship (Phase 2). Participants were
137 not given a score or rank for their second spaceship at the end of Phase 2, meaning that they
138 received no feedback on whether the choice to use asocial or social source of information

139 improved the outcome. Furthermore, because scores were randomly generated, no rules for
140 building high-scoring spaceships were available for participants to learn.

141 Phases 1 and 2 were then repeated a further two times (i.e., three Rounds of building
142 spaceships), with scores and ranks shown at the end of each Phase 1. Each participants'
143 spaceships were ranked randomly, once at 1st, 3rd or 5th on the league table. At the end of
144 each Round, participants were informed that their score had been saved and that their best
145 score out of the three Rounds would be used at the end of the experiment to allocate the bonus.
146 Because the scores given to spaceships were random, bonuses were awarded at the end of the
147 experiment to more than one participant according to a lottery in which participants had a higher
148 probability of a reward when choosing the safe rather than the risky option in the final Round.

149 After completing all three Rounds, participants completed the 12-item risky impulsivity
150 measure [28] on-screen. This measure assesses willingness to take risks without prior thought
151 in everyday life and is reported to have high internal consistency [28]. The bonus payments
152 were awarded when all participants had completed the final on-screen material, including the
153 questionnaire.

154

155 *Asocial and social options*

156 The asocial option consisted of viewing up to ten previously unseen items in the scrapheap, of
157 which up to three items could be kept for use in the next building Phase. The social option
158 consisted of viewing three completed spaceships, ostensibly built by 'other participants', along
159 with the associated 'scores'. These spaceship designs had actually been generated by the
160 experimenter prior to the study using randomly selected tiles, and three out of twelve completed
161 spaceships were presented at random as social sources. The scores for these spaceships were
162 also randomly assigned. Participants could choose up to three items from one of the three
163 ships, and these items were automatically added to the participant's spaceship template in the

164 next Phase and could not be removed. Because each participant was assigned to a single
165 condition, the description of the asocial and social options (see below) remained the same for
166 participants across Rounds, in order to avoid potential confusion among participants and reduce
167 the chance that participants failed to attend to the subtle differences in the descriptive material.

168 In the Asocial Risky (AR) condition (N=28; 18 female, 10 male), participants were
169 informed that their score could markedly increase or decrease if they visited the scrapheap, in a
170 short paragraph that included the following wording: “some of these items may be broken and
171 useless, but some may greatly increase your ship’s score... your score could go up or down”.
172 Conversely, the social option was safe; participants were informed their score could slightly
173 increase or would remain the same (“the ships will have the same score as your ship, or slightly
174 higher... you will be guaranteed at least the same score as your current ship”). In the Social
175 Risky (SR) condition (N=32; 17 female, 15 male), the asocial option was safe (“all of these items
176 will help your ship to fly, and some of them can slightly increase your ship’s score... you will be
177 guaranteed at least the same score as your current ship”), and the social option was risky (“the
178 ships may have a much worse or much better score than your current ship’s score... your score
179 could go up or down”). In the Control (C) condition (N=28; 15 female, 13 male), both the asocial
180 and social options were safe: the wording was identical to that used in the safe options in the
181 other conditions. This wording reflects the Bounded Risk Distribution model, in which individuals
182 are expected to maximize their probability of reaching a goal while minimizing their probability of
183 falling below a certain threshold [27]. In our experiment, participants are trying to achieve the
184 goal of a top score and want to minimize their chance of falling below this threshold, in order to
185 achieve a bonus payment. Therefore, although the safe options have a slightly higher average
186 expected score, these safe options preclude a large increase in score. The risky option is
187 therefore a rational choice where participants believe that they need to greatly improve their
188 score in order to move up in the rankings and win a monetary bonus. .

189

190 *Statistical analyses*

191 We modelled the participants' decision to use asocial or social options using Bayesian binomial
192 multi-level logistic regression in R with the *map2stan* function from the *Rethinking* package [29].
193 The full model included an effect for sex, an effect for condition, a sex*condition interaction, an
194 effect for the rank given to the participant's spaceship after Phase 1, and a random effect for
195 individual. The C condition was represented as the baseline in the model, so that any effects of
196 the AR or SR conditions were in relation to C. Because men were coded as 0 and women as 1,
197 the baseline represents men's behaviour in the control condition, and the effect of sex
198 represents how women's behaviour differed from men's in the C condition. Model predictions
199 were calculated by averaging across all candidate models weighted according to the WAIC
200 (Watanabe-Akaike Information Criteria). The model with the lowest WAIC value, and the highest
201 Akaike weight, is the model that is most likely to make accurate predictions on new data,
202 conditional on the set of models considered. Posterior predictions were calculated based on the
203 population mean of the participants and thus represent predictions for a 'new', previously
204 unobserved, average participant. These predictions are presented in Figure 2. Candidate
205 models were chosen based on *a priori* hypotheses formulated before data collection (**Table 1**).

206 In order to examine whether the choice of using risky or safe options varied with the rank
207 assigned to the spaceship, or sex of participant, and whether men and women responded
208 differently to their rank assignments, we ran an additional model with risky/safe choice rather
209 than social/asocial choice as the outcome variable. This model excluded data from the C
210 condition, because both options in this condition were safe and therefore no risky choice could
211 be made. The risky/safe choice was modelled using a Bayesian binomial multi-level logistic
212 regression with rank, sex and a sex*rank interaction as predictors.

213 Finally, we also modelled participants' risky impulsivity scores using a Bayesian linear
214 model, with sex as a predictor variable, to check whether our sample displayed the expected

215 sex difference in risky impulsivity.

216 All model estimates are reported with 89% credible intervals (CIs), which are the default in
217 the *Rethinking* package [28]. The CIs provide an upper and lower estimate around the mean of
218 the parameter estimate and encompass 89% of the posterior distribution. This method contrasts
219 with the traditional use of 95% confidence intervals in null hypothesis testing. Using 95%
220 intervals would not change the interpretation of our results, because we are using a model
221 comparison approach, and the size of the credible intervals does not affect which models best fit
222 the data. All error bars are 89% credible intervals and can be interpreted as the region within
223 which the model expects to find 89% of responses, given the data and the assumptions in the
224 model.

225

226 Results

227 *Asocial versus social options model*

228 When modelling the probability of choosing asocial or social options, the best-fitting model (i.e.,
229 the model with the lowest WAIC value) included an effect for sex and an effect for the sex and
230 AR condition interaction (**Table 1**). This interaction can be seen in detail in **Figure 3**. In the C
231 condition, both women and men preferentially chose to use the asocial source information
232 rather than the social source (women: $\beta = -0.72$, CI [-1.36, -0.05]; men: $\beta = -0.41$, CI [-0.72, -
233 0.10]; **Figure 3**). As shown by the model estimates (**Figures 4 & 5**), there was not strong
234 evidence for an interaction effect between sex and SR condition ($\beta = 0.62$, CI [-0.09, 1.40]),
235 meaning that women's choices in the SR condition did not differ strongly from women's choices
236 in the C condition. Thus, as can be seen in the model predictions (**Figure 3**), both women and
237 men preferentially chose the asocial source in the SR condition also. In contrast, the interaction
238 between sex and the AR condition had a strong effect in the model ($\beta = 1.76$, CI [1.03, 2.51];

239 **Figure 4**). As can be seen in **Figure 4**, women in the AR condition preferentially chose the
240 social option, whereas men's choice did not differ compared to men's choices in the control
241 condition. Thus, women in the AR condition chose the social source of information more than
242 women in the C condition, while men in the AR condition did not differ from men in the C
243 condition with regard to their choice. According to the full model (**Figure 5**), rank did not predict
244 the choice to use asocial or social options ($\beta = 0.05$, CI [-0.22, 0.29]).

245

246 *Risky versus safe model*

247 The risky versus safe model indicated that participants of both sexes preferred to use the safe
248 option overall ($\beta = -2.38$, CI [-4.20, -0.56]; **Figure 6**). The intercept estimates the preferences of
249 men in both conditions, showing that they had an overall preference for the safe choice, and the
250 effect of sex included zero ($\beta = 1.12$, CI [-0.71, 3.13]), indicating that women did not choose
251 differently from men. However, the effect of rank ($\beta = 1.14$, CI [0.22, 1.95]) shows that both men
252 and women were more likely to choose risky than safe options after receiving a lower rank than
253 a higher rank. There was no evidence for an interaction between rank and sex in the model ($\beta =$
254 -0.77 , CI [-1.98, 0.19]), indicating that men and women were responding similarly to their rank
255 assignments.

256 *Risky impulsivity measure*

257 Women scored lower than men on the risky impulsivity measure, as expected (women = $23.41 \pm$
258 6.97 ; men = 27.24 ± 7.59 ; means and SEMs) ($\beta = -0.06$, CI [-0.09, -0.04]; **Figure 7**). Men
259 scored half a standard deviation higher than women on average (Cohen's $d = 0.52$). The scale
260 had an acceptable level of internal consistency (Cronbach's $\alpha = 0.78$). Individual scores did
261 not predict the use of risky versus safe options ($\beta = 0.13$, CI [-1.75, 1.90]).

262

263 **Discussion**

264 In our experimental study, we found a sex difference in the use of social sources of information
265 that only emerged when using the asocial option was a risky (i.e. the payoffs varied widely).
266 Women, but not men, preferentially chose to use the social source of information when the
267 asocial option was risky. In contrast, women and men did not differ from each other in their
268 responses to risky social options; both women and men preferentially used the asocial option in
269 the 'Social Risky' condition, as well as in the control condition. Male and female participants
270 were more likely to choose the risky option when the spaceship was given a low rank than a
271 high rank, while rank did not predict whether participants chose asocial or social information.
272 Women had lower average scores than men on the risky impulsivity measure, as reported in
273 previous research (e.g., [28, 30]). Our main finding, which was that individuals of the more risk-
274 averse sex (i.e. women) used the social option when the asocial one was risky, is consistent
275 with the hypothesis that levels of risk-aversion influence social learning strategies [4]. This result
276 has potentially broad implications for understanding the dynamics of social information
277 transmission.

278 While previous research has suggested that women are more likely than men to
279 conform to the decisions of others (e.g., [31]), our findings contribute further evidence that social
280 sources of information are used strategically, irrespective of gender. We found that the sex
281 difference in the use of social sources of information depended upon the type of decision being
282 made. Women were not more likely than men to use social options across all conditions, nor
283 were women less likely than men to choose the risky option in general. The sex difference in the
284 use of the social sources of information when the asocial source was risky could potentially
285 have reflected lower confidence in one's own performance in women compared to men.
286 Previous research has shown that both female and male participants copy others when lacking
287 confidence in their personal information (e.g. [31]), and that this relationship is likely to influence

288 patterns of conformity in cases where men's and women's confidence differs [32]. However, the
289 absence of a sex difference in the control condition suggests that both sexes were equally
290 confident in solving the task alone.

291 The psychological mechanisms underpinning the sex difference in response to risky
292 asocial sources remain to be determined. While a sex difference in competitiveness has been
293 identified in previous literature [33], this sex difference is unlikely to explain our results because
294 men and women responded similarly to their Phase 1 ranks. One possible explanation is that
295 women were more sensitive on average than men to the potential loss in score associated with
296 the risky asocial option and were thereby minimizing their probability of a loss. However, women
297 and men did not differ in their probability of selecting a risky social option, possibly because they
298 had a preference for the asocial option irrespective of risk. Neither female nor male participants
299 avoided the social option completely, even when it was risky. Participants might have been
300 sampling the social sources in order to compare their own decisions with those of others or to
301 check for particularly high-scoring solutions. This sampling strategy might have prevented
302 participants from relying solely on the asocial option in the social risky condition, which might
303 have resulted in a ceiling effect. Altering the experimental design to make the social option more
304 appealing (in terms of perceived benefits gained from viewing social sources) might have
305 reduced overall reliance on asocial sources when this social information became risky

306 Our results showed that both men and women used asocial, rather than social, sources
307 of information when both sources were safe, consistent with previous experiments showing the
308 preferential use of asocial learning in laboratory settings (e.g., [32, 34]). While theoretical
309 models have suggested that social learning should initially be prioritised over asocial learning
310 [35], our empirical research suggests that participants prefer to try to solve tasks for themselves,
311 before relying on help from others. The asocial version of our task, which involved viewing new
312 tiles in a scrapheap, could have been more appealing than the social condition, in terms of
313 providing opportunities to innovate or for other reasons related to the characteristics of the

314 stimuli. Although the probability that men chose the social option did not vary across conditions,
315 the level of risk could potentially influence use of social sources of information by men under
316 different experimental conditions. For instance, further increasing the riskiness of the asocial
317 option could potentially result in men switching to using the social option. The idea that men and
318 women could differ in the cut-off point at which the risk is deemed sufficiently high to change
319 strategy could be investigated experimentally by varying the level of risk along a continuum.

320 Our results confirmed that participants of both sexes were more likely to choose the risky
321 option when the spaceship was given a low rank than a high rank. Previous experiments using
322 economic game protocols have also shown that participants are more likely to take a risk when
323 performing poorly in relation to other participants (e.g., [37]). Given that our definition of
324 riskiness focused on the variation in expected score, rather than the absolute size of the
325 expected score, future studies could manipulate both average scores and variance in scores of
326 different options to examine the influence of these on the strategic use of social information in
327 men and women. In the current study, the average score for the safe option was slightly higher
328 than for the risky option, given that scores in the safe option could either increase slightly or
329 remain stable, while scores in the risky option could either increase or decrease markedly. The
330 effects of manipulating level of risk and differences in average payoffs could be further
331 investigated experimentally in both human beings and non-human animals, using available
332 protocols (e.g., [7, 10, 19]).

333 Individual scores on the risky impulsivity measure did not correlate with the likelihood of
334 choosing the risky versus safe option. While this correlation was not the main focus of the study,
335 one suggestion for future research might be to examine measures of sensitivity to 'actuarial'
336 risk, although these measures are less likely to show sex differences than measures of
337 sensitivity to physical risk [20]. Indeed, we chose risky impulsivity because we were looking for a
338 trait that differs by sex. A second possibility is as follows. In our task, choosing the 'risky' option
339 is rational when a large increase in score is needed, but it brings with it the possibility of a large

340 'loss' in score. We could consider a decrease in score when selecting the rational option as a
341 form of unrepresentative negative feedback [38, see also 26], to which women appear to be
342 more sensitive than men. We might therefore expect sensitivity to negative feedback in, for
343 example, a gambling task to correlate with a shift in strategy in our spaceship-building task.

344 **Conclusion**

345 Our results indicated that individuals of the more risk-averse sex preferentially used a
346 social option when the asocial option was risky, supporting theoretical evidence that levels of
347 risk-aversion are linked to the implementation of social learning strategies [4]. Whether the
348 psychological mechanisms underpinning the decision to use social sources of information
349 involved greater sensitivity to punishment or lower confidence in one's own performance was
350 not investigated. However, regardless of the mechanism, switching to social learning can
351 potentially provide individuals with the opportunity to avoid costly mistakes and learn from the
352 successes of others. Understanding how sex differences in risk-aversion relate to the use of
353 social information deserves further investigation in non-human animals, as well as humans, and
354 would add to the growing evidence that individual traits influence a broad range of social
355 processes [12, 40]. Between-individual differences in risk-aversion are likely to influence the
356 dynamics of social learning and the spread of socially transmitted information through
357 populations, with potential broad-scale implications for the characteristics of local traditions and
358 the evolution of cultural traits.

359

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366

367 **Supplementary information**

368 The full dataset supporting this paper is available at [https://risweb.st-](https://risweb.st-andrews.ac.uk/admin/workspace.xhtml?uid=5)
369 [andrews.ac.uk/admin/workspace.xhtml?uid=5](https://risweb.st-andrews.ac.uk/admin/workspace.xhtml?uid=5)

370

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478 **Figure Legends**
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Table 1 (on next page)

Candidate models and their WAIC weights

List of candidate models that were included in the asocial/social information analysis, the a priori hypotheses and the included parameters, with model values (WAIC \pm SE) and weights (Akaike weight). Bold type indicates the best fitting model

1
 2 **Table 1** List of candidate models that were included in the asocial/social information analysis,
 3 the a priori hypotheses and the included parameters, with model values (WAIC \pm SE) and
 4 weights (Akaike weight). Bold type indicates the best fitting model.

5

Model	Hypothesis	Parameters included	WAIC (\pm SE)	Akaike weight
1	Null	Intercept	360.3 (5.10)	0.00
2	Full	Intercept + sex + AR + SR + sex*AR + sex*SR + rank + personality	357.1 (10.36)	0.01
3	Sex and condition interactions predict choice	Intercept + sex*AR + sex*SR	351.1 (9.11)	0.25
4	Sex, and sex and condition interactions, predict choice	Intercept + sex + sex*AR + sex*SR	349.6 (9.77)	0.52
5	Sex and condition predict choice	Intercept + sex + AR + SR	354 (8.92)	0.06
6	Only condition predicts choice	Intercept + AR + SR	352 (8.77)	0.16
7	Only sex predicts choice	Intercept + sex	362.3 (5.32)	0.00

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7














Figure 1

Screenshot of Phase1

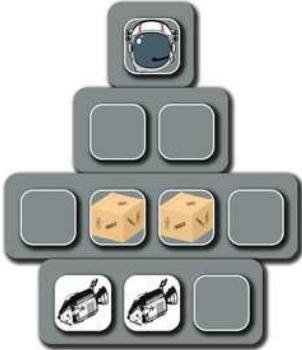
Example screenshot from the online experiment showing shipbuilding in Phase 1.

You are on Phase 1 of ROUND 1

Perishable foodstuffs

Crew	<input type="checkbox"/>				<input type="checkbox"/>	<input type="checkbox"/>
Cargo	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>			<input type="checkbox"/>
Engine	<input type="checkbox"/>			<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Shields	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>			
Lazers	<input type="checkbox"/>	<input type="checkbox"/>				<input type="checkbox"/>

REMEMBER:
You have to use at least one engine and one crew member – and you only have TWO MINUTES!



Time Remaining: 01:11

Figure 2

Screenshot of Phase1.

Example screenshot of the generated score presented to participants at the end of Phase 1.

You have completed Phase1

Well done! Your ship's score is 25,000. Here is a list of the high scores:

Name	Score
Rank 1	25,000
Rank 2	21,050
Rank 3	16,850
Rank 4	10,500
Rank 5	8,780

Click continue to find out how you can improve your score!

Continue

Figure 3

Model predictions

Model predictions of the mean proportion of individuals that chose social information plotted according to condition and sex. Predictions were averaged across all models and weighted according to WAIC weight. Error bars show 89% CIs. Raw means are also plotted, represented by a cross symbol

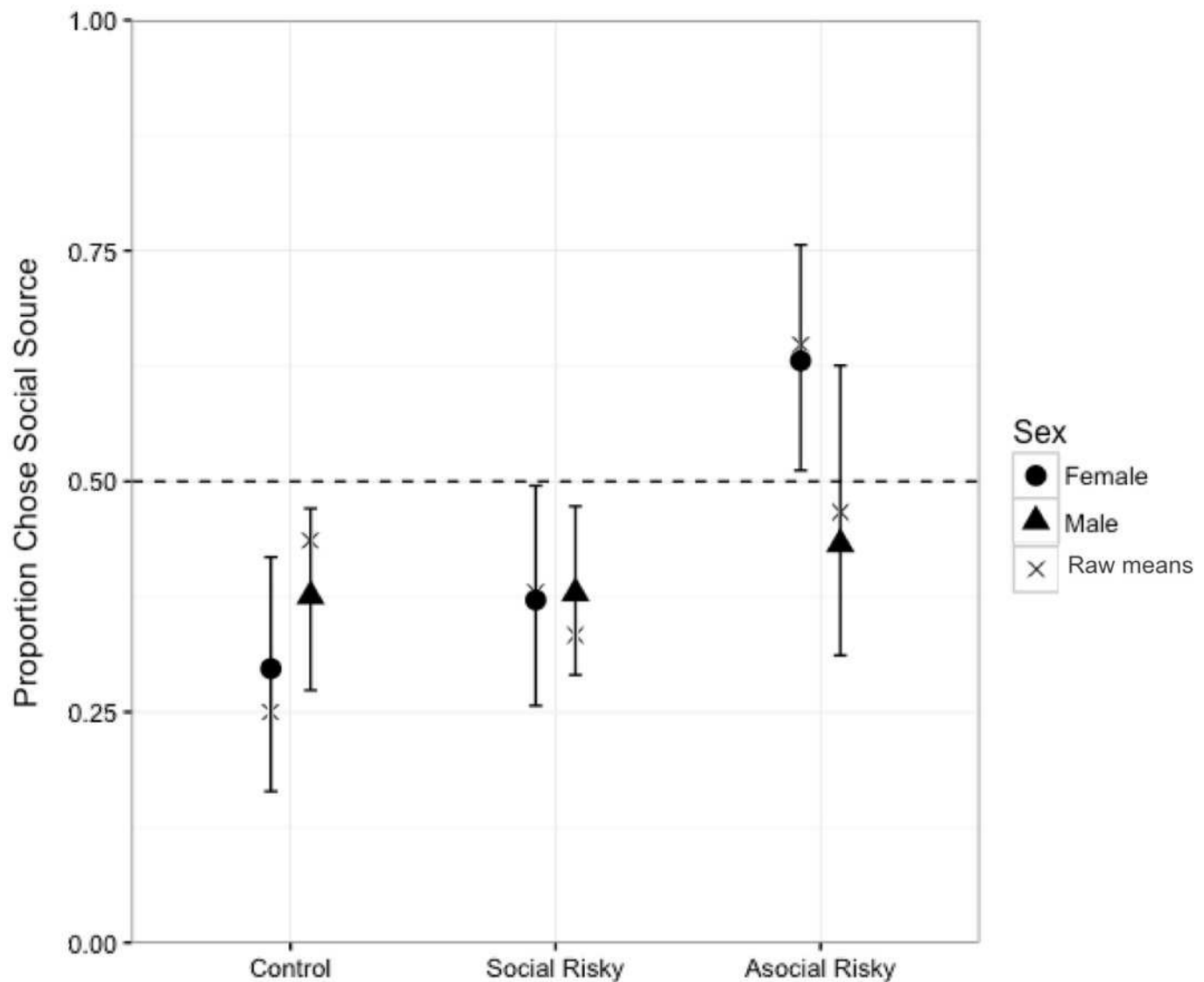


Figure 4

Plot of parameter estimates of best fitting model (lowest WAIC)

Plot displaying parameter estimates for the probability of choosing the social option, taken from the model with the lowest WAIC value and plotted with 89% CIs. A positive estimate indicates a greater likelihood of choosing social, rather than asocial, information. Where the 89% CIs of parameter estimates include zero, there is no clear evidence of an effect of that parameter on the likelihood of choosing the social or asocial option. The intercept (baseline) represents males in the control condition

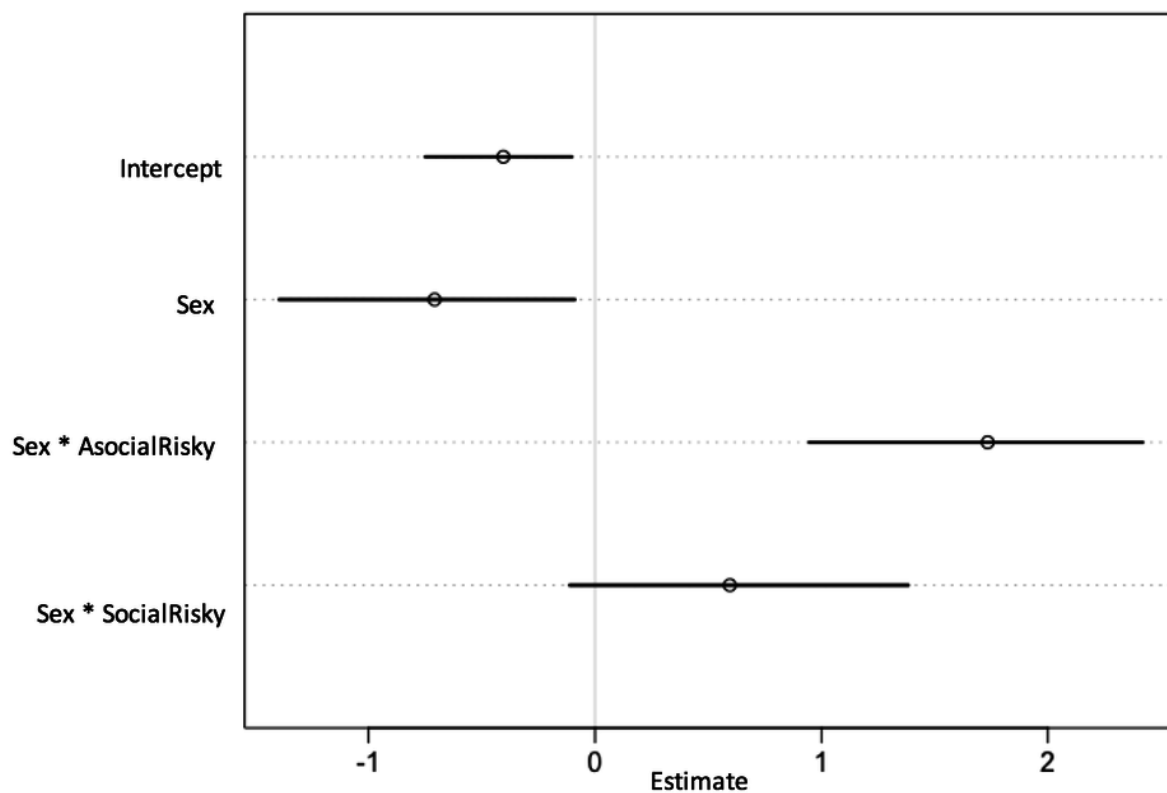


Figure 5

Plot of parameter estimates from Full Model.

Plot displaying parameter estimates for the probability of choosing the social option, taken from the full model and plotted with 89% CIs.

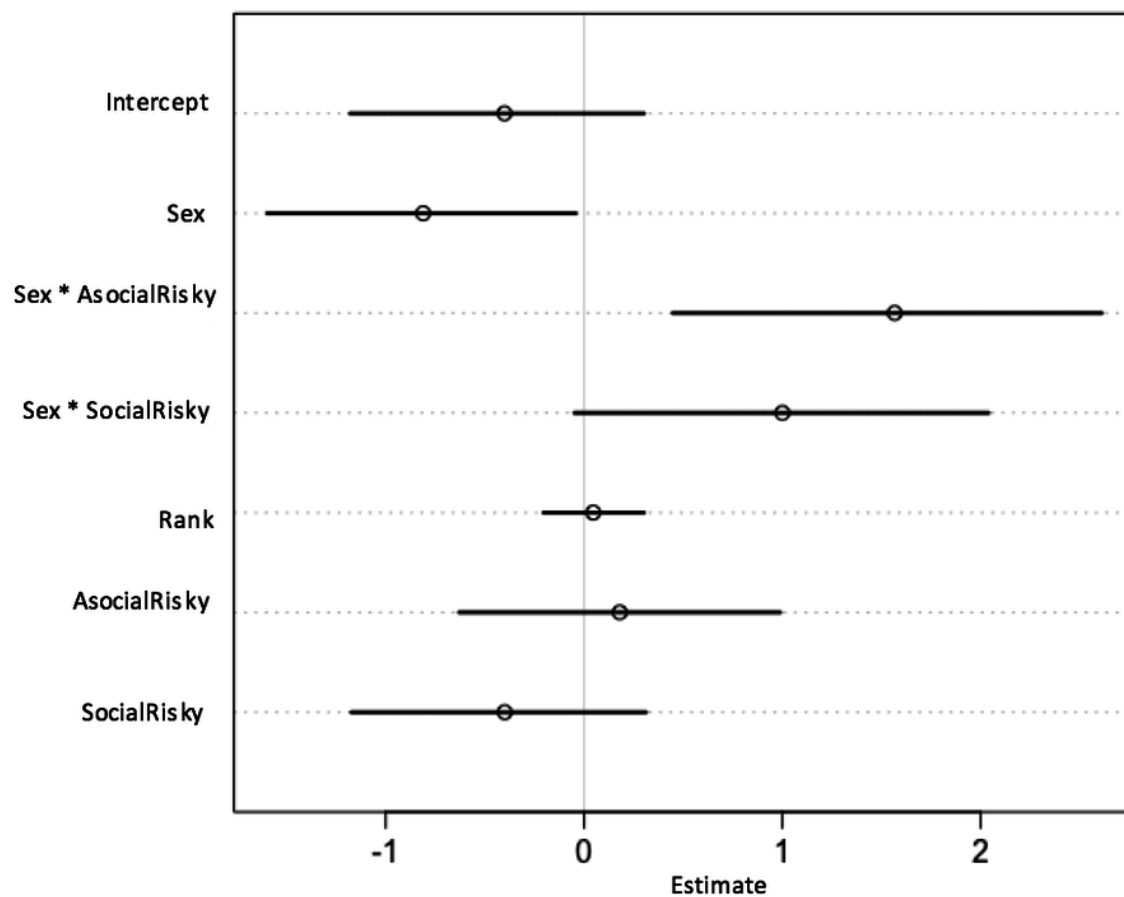


Figure 6

Plot of parameter estimates from the second model.

Plot displaying parameter estimates for the likelihood of choosing the risky option, plotted with 89% CIs. A positive estimate indicates a greater likelihood of choosing the risky, rather than the safe, option. Where estimates include zero, there is no clear evidence of that parameter affecting the likelihood of choosing the risky or safe option

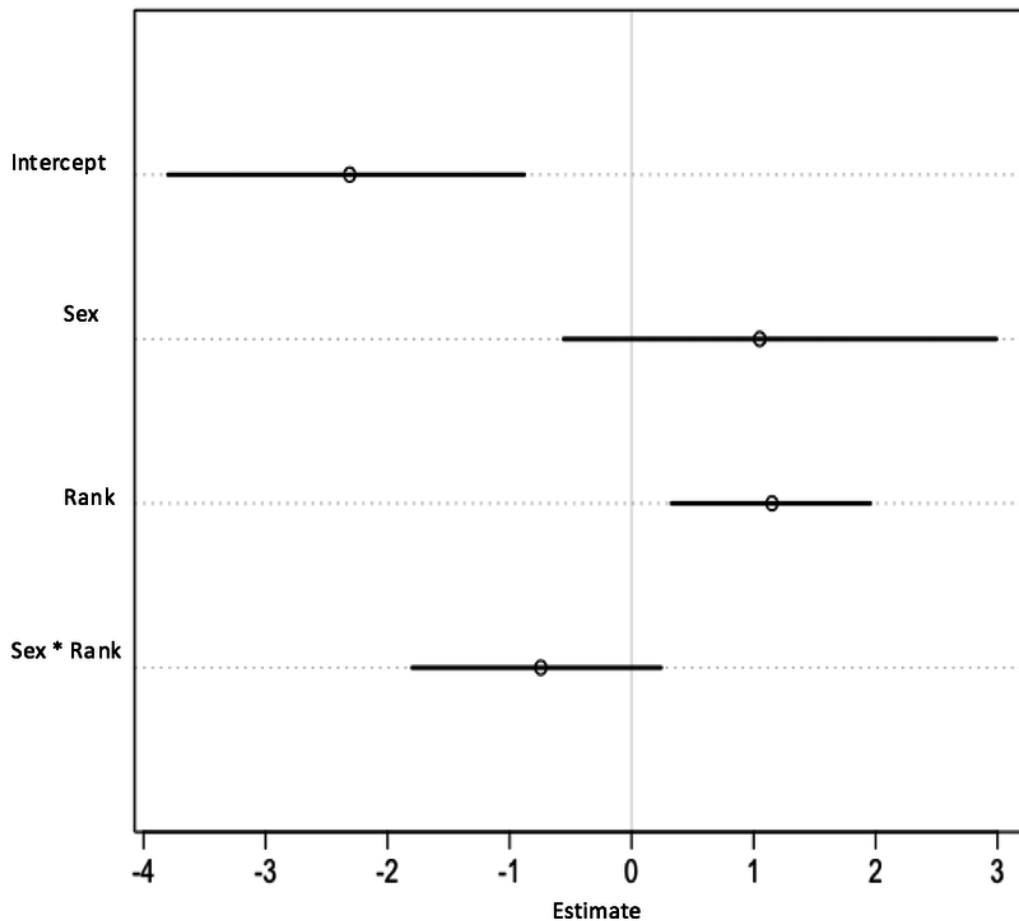


Figure 7

Density plot of sex difference in risk-taking measure

Density plot showing men and women's risky impulsivity scores

