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The illusion of control: a Bayesian perspective

Adam J. L. Harris · Magda Osman

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Abstract In the absence of an objective contingency, psychological studies have shown that people nevertheless attribute outcomes to their own actions. Thus, by wrongly inferring control in chance situations people appear to hold false beliefs concerning their agency, and are said to succumb to an illusion of control (IoC). In the current article, we challenge traditional conceptualizations of the illusion by examining the thesis that the IoC reflects rational and adaptive decision making. Firstly, we propose that the IoC is a by-product of a rational uncertain judgment (“the likelihood that I have control over a particular outcome”). We adopt a Bayesian perspective to demonstrate that, given their past experience, people *should* be prone to ascribing skill to chance outcomes in certain situations where objectively control does not exist. Moreover, existing empirical evidence from the IoC literature is shown to support such an account. Secondly, from a decision-theoretic perspective, in many consequential situations, underestimating the chance of controlling a situation carries more costs than overestimating that chance. Thus, situations will arise in which people will incorrectly assign control to events in which outcomes result from chance, but the attribution is based on rational processes.

Keywords Bayesian · Illusion · Control

The Illusion of Control (IoC) indicates a perceived causal relationship between actions and outcomes in the absence of an objective contingency. Moreover, people not only

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20 attribute responsibility to desirable chance outcomes, but also undesirable ones (Biner
21 et al. 2009; Langer 1975; Mirowsky and Ross 1990). In processing terms, this implies
22 that we either encode an association between our actions and observed outcomes
23 (when none exists), or incorrectly infer that through our actions we can influence
24 a chance outcome. The prevailing opinion is that, while objectively there is zero
25 contingency between actions and outcomes in some contexts (e.g., gambling scenar-
26 ios, experimental settings), in most other contexts judging control is important from an
27 adaptive perspective (Blanco et al. 2011) and may even be innate (Leotti et al. 2010).
28 However, laboratory demonstrations of the IoC are taken as evidence that we make
29 false attributions of causality, and maintain erroneous beliefs of control. In this article
30 we propose that the IoC may be usefully situated within a Bayesian framework.

31 We follow a long tradition in assuming that rationality criteria should be based on
32 information that is available to the agent. For example, if there are 20 red balls in
33 an urn and 80 black balls, predicting that the next ball will be black is the rational
34 decision regardless of the actual outcome. Participants are said to display the IoC if
35 they perceive control in a situation where, objectively, there is none (i.e., the bias is
36 evaluated on a privileged external standard). Here we assess the information available
37 to the individuals who display the IoC. We argue that the IoC is a necessary by-product
38 of a rational judgment (on standards of Bayesian probability and Decision Theory),
39 given the statistics of the world in which we live.

40 In the present article, we do not purport to provide an exhaustive review of the IoC
41 literature. Rather, we propose a framework which links IoC research to ‘big’ ques-
42 tions concerning the fundamental representational structure of the cognitive system
43 (e.g., Chater et al. 2010, 2006), as well as ‘big’ issues concerning ways in which we
44 reduce uncertainty. Firstly, we consider a Bayesian approach as a way of identifying
45 in a principled manner what makes a situation controllable. The approach prescribes
46 that, when making a judgment about the controllability of a situation, agents should
47 combine the available evidence with their prior degree of belief. In support, there is
48 evidence demonstrating that cues typically more representative of controllable than of
49 uncontrollable tasks are reflected in judgments of control in laboratory based IoC tasks
50 (e.g., Langer 1975; Langer and Rodin 1976). Moreover, we propose that people’s prior
51 degrees of belief should reflect the fact that most naturalistic scenarios are controllable.

52 Following judgments as to the likelihood that a situation is controllable, the next
53 step is to decide whether to act as though the situation is controllable or uncontrollable.
54 We propose Decision Theory as the framework within which such a decision should
55 be made. The decision will thus be influenced by the utilities associated with misclas-
56 sifications of controllable events as uncontrollable (‘misses’/the ‘illusion of chaos’)
57 and uncontrollable events as controllable (‘false positives’/the IoC). We propose that
58 asymmetries in the utilities associated with these different errors further enhance the
59 likelihood for an objectively uncontrollable situation to be rationally, although wrongly
60 (from an objective perspective), classified as controllable (the IoC).

61 1 Bayesian rationality

62 The main sources of evidence for the IoC have originated from artificial environments
63 (e.g., Psychology Laboratories, Casinos), which have been deliberately constructed

64 so as to give rise to an illusory perception of control. The first argument we make
 65 is that participants in IoC experiments optimally use the evidence available to them
 66 when making judgments of the controllability of a situation. Thus, the IoC is a result
 67 of rational Bayesian updating. Psychological experiments investigating the IoC are
 68 unusual in that the conditions imply control when, objectively, control is absent.

69 Within the Bayesian framework, degrees of belief are represented as single event
 70 probabilities (or distributions over these probabilities). Here we propose that people
 71 have a degree of belief in the truth of the hypothesis ‘Outcome X is controllable’.
 72 Bayes’ Theorem prescribes how people should update their degree of belief in a
 73 hypothesis (e.g., that an outcome is under their control) upon receipt of a piece of
 74 evidence:

$$75 \quad P(h|e) = \frac{P(h)P(e|h)}{P(e)}$$

76 Thus, a person’s posterior degree of belief in the hypothesis, given a piece of
 77 evidence, $P(h|e)$, is a function of their degree of belief in the hypothesis before
 78 receiving evidence (the *prior*), $P(h)$, the likelihood of the evidence given the truth
 79 of the hypothesis, $P(e|h)$, and the base rate of the evidence, regardless of the truth
 80 status of the hypothesis, $P(e)$. The main support for our proposal, that the IoC is a
 81 result of such rational belief updating, comes from the evidence people appear to use
 82 to make inferences about controllability in IoC studies. In the remainder of this article,
 83 we provide evidence in support of our first argument, that the IoC results from rational
 84 belief updating, by considering the role of evidence (e.g., cues to controllability).
 85 Secondly, we discuss the limited available evidence concerning the role of priors, and
 86 propose this as a fruitful direction for further research. Finally, we outline the decision
 87 theoretic framework within which we locate our account.

88 2 The role of evidence

89 [Langer \(1975\)](#) claimed that people conflate chance- and skill-based situations because
 90 choice and active-involvement in a chance based task were the critical factors that
 91 generated the IoC. We consider this popular interpretation of the IoC from the point of
 92 view that the conflation between chance and skill is not erroneous, rather people are
 93 inferring cues to skill from IoC tasks (i.e., updating their degree of belief that a task
 94 is controllable from the supporting evidence), and basing their actions on this. From
 95 a Bayesian perspective this is a rational basis for judging control.

96 2.1 Choice and skill

97 From early demonstrations of the IoC and associated work on control (e.g., [Langer](#)
 98 [1975](#); [Langer and Rodin 1976](#)), to recent studies on choice and gambling ([Martinez](#)
 99 [et al. 2009](#)), having a choice within a task has been shown to increase perceptions of
 100 control. The relationship between choice-behaviors and the IoC has therefore been
 101 a popular basis for understanding the illusion. To gain insight into this relationship,

we turn to situations which involve choice and chance, such as gambling scenarios (e.g., a roulette player can *choose* to bet on red or black, but this does not change her probability of winning [*chance*]). Gambling scenarios, or variations of them, have dominated IoC research. The rationale being that people's underlying motivation is to act in ways that will increase their success in achieving desirable outcomes. The probability of desirable outcomes will increase as a result of choice in skill-based tasks. In chance-based situations we therefore tend to make a misattribution that we can influence the outcomes when there is some choice in the tasks used. We know that there are several factors associated with gambling contexts which help to heighten our perception of agency over future outcomes (Burger 1986; Langer 1975; Langer and Roth 1975). For instance, they are often interactive which means that not only are our choices incentivized, but we are invested in the game because we directly act in it and can become familiar with it. Clearly it is wrong to hold beliefs of agency in zero contingency contexts, but it may not necessarily be irrational.

As outlined in the 'Bayesian Rationality' section above, to inform our judgment, we should combine our prior with the evidence at our disposal. The cues (e.g., choice and familiarity) that typically result in a greater likelihood of the IoC can be considered as items of evidence (Langer 1975), and thus our belief in the controllability of the situation *should* increase. For instance, amongst many of the factors that help improve our skill in performing cognitive and motor tasks (e.g., playing the piano, flying a plane, operating the latest mobile phone handset) is building up familiarity with the task through extensive practice (e.g., Ericsson and Lehmann 1996). When providing participants with repeated choices in IoC tasks, it is therefore no surprise that people are willing to attribute rewarding outcomes to their own agency, because the task conditions provide evidence that skill can influence the outcome (Langer 1975).

2.2 Active involvement

The proposal that participants use evidence from the task to make judgments about the controllability of the scenario is not a new argument, nor does it imply a Bayesian explanation. Matute's (1996; see also, Blanco et al. 2011) framing of the IoC within an associative learning approach specifies the role of active involvement in the IoC in terms of the probability of responding (interacting with the task), $P(R)$. On this account, the IoC is largely driven by the frequency of occasions on which an individual acts in an environment. Differences in the number of times a response is made leads to differences in the data/evidence that is obtained, consequently the IoC can arise in simulations of a standard associative learning model, such as the Rescorla-Wagner model (Matute et al. 2007). In order to most accurately perceive the degree of contingency between an action and an outcome, an agent should only make an action on half the available trials, so as to understand the number of times in which the outcome occurs in the absence of the action, as well as in the presence of the action. As the frequency of responding increases, so will the exposure to spurious associations between actions and positive outcomes; because responses appear to be reinforced this will further increase the probability of responding in the future (Blanco et al. 2011). Blanco et al. (2011) task, a variant of Alloy and Abramson (1982) learned helplessness study,

145 involved the presentation of 50 fictitious medical case notes from patients with the
 146 same rare disease. Participants had to decide whether they would administer medicine
 147 (by pressing a button) or not (not pressing a button). They then received feedback on
 148 the status of the patient (“well”, “still ill”) but crucially there was zero contingency
 149 between the choice to medicate and the patient’s status of health. As the number of
 150 responses (button presses) increased, so too did the estimates of success in generating
 151 a positive outcome (IoC), regardless of the actual, objective, contingency.

152 Work in the associative learning domain suggests that evidence (i.e. outcomes asso-
 153 ciated with repeated actions) influences the degree to which people succumb to the
 154 IoC. Critically, however, the way in which evidence is experienced (i.e. through active
 155 involvement) is, on an associative account, crucial to generating the IoC. However,
 156 choice appears to influence the IoC above and beyond repeated active response in
 157 chance based situations, for example the IoC can arise from a single exposure to a
 158 choice event (Langer 1975; Martinez et al. 2009). In this case the IoC is not dependent
 159 on multiple experiences of action-outcome associations. The associative account rests
 160 on demonstrations of the IoC in situations in which people have repeated exposure
 161 to probable reinforcers or repeated active involvement in a chance based situation.
 162 This account, however, is clearly limited given the range of conditions in which the
 163 IoC phenomena has been reported. The Bayesian account is proposed as a unifying
 164 framework within which all instances of the IoC can be understood.

165 In addition to highlighting the role of evidence, the Bayesian account also stipu-
 166 lates that evidence should be combined with *priors*, a consideration that cannot be
 167 incorporated within an associative learning account.

168 3 Priors

169 Recall our earlier example of an urn filled with 20 red balls and 80 black balls. By
 170 extension, if, over the course of your lifetime, you have had many exposures to urns
 171 and the sizeable majority contained a preponderance of black balls, then before draw-
 172 ing any balls from the new urn your prior expectation will be that there are more black
 173 balls than red balls. We propose that over the course of a person’s lifetime, most people
 174 perceive more situations that are controllable than situations that are uncontrollable.
 175 Consequently, upon entering an experimental situation, especially one with cues that
 176 are reminiscent of controllable situations (e.g., skill, choice, active involvement), your
 177 expectation (i.e., prior) is that this is more likely to be a situation that is controllable
 178 than one that is uncontrollable ($P[\text{controllable}] > .5$).

179 Despite empirical evidence for the IoC, to date there are no dedicated investigations
 180 of how prior experience influences judgments of control. Such studies need to explore
 181 the differences in people’s expectancies (priors) for different types of contingency in
 182 different experimental situations. Only by considering all possible contingency rela-
 183 tionships can the role of priors be completely understood, because some situations
 184 will invoke prior beliefs of a positive contingency, others a negative contingency, and
 185 still others, no contingency. To explore the role of priors, a focused empirical research
 186 program would therefore systematically manipulate priors, before presenting partic-
 187 ipants with a situation in which either a positive, negative or no contingency exists.

188 Without such focused and controlled research, interpretations of previous results that
189 might be perceived as indirectly investigating this issue are speculative. Nevertheless,
190 in this section we aim to highlight the key role that priors are presumed to play in the
191 IoC, and present a possible fruitful line of research on the IoC that could be explored.

192 To make sense of findings from the IoC literature [Mirowsky and Ross \(1990\)](#) have
193 also suggested a role for the priors that people bring to an experimental situation from
194 their life experience. The effect that they focus on is the commonly reported finding
195 that the IoC is observed less in depressed individuals than in non-depressed individuals
196 (e.g., [Alloy and Abramson 1979](#)).

197 [Mirowsky and Ross \(1990\)](#) examined the effects of individuals' judgments of
198 responsibility for good/bad events, and good/bad outcomes. Non-depressed individ-
199 uals gave high ratings of control over all four types of events and outcomes, whereas
200 depressed individuals judged all four types of events as out of their control. These
201 differences were explained on the basis of the locus of control (LoC). When the LoC
202 is judged to be internal, it is associated with other aspects of behavior that indicate
203 high levels of control (e.g., motivation, goal-directed thinking, ambition). When the
204 LoC is judged to be external, in other words external factors are thought to influence
205 outcomes (e.g., luck, fate, chance, powerful others, unpredictable complex mecha-
206 nisms), judgments of control are lower for *all* events, controllable and uncontrollable.
207 This explanation has since been supported by findings from [Stadelhofen et al. \(2009\)](#),
208 who examined the relationship between LoC and depression in patients with path-
209 ological gambling problems. In addition, those scoring high in desire for control, a
210 scale that examines the extent to which individuals are motivated to control outcomes
211 in their environment ([Burger and Cooper 1979](#)), tended to make higher attribution
212 judgments of negative as well as positive outcomes, as compared to those scoring
213 low on the scale ([Burger and Hemans 1988](#)). [Mirowsky and Ross](#) argued that actually
214 the real probabilities of a win or a loss of a simple laboratory task is not what deter-
215 mines the IoC. Rather, it is the prior beliefs of the probabilities of wins and losses that
216 people bring with them to the lab. Depressed participants may have had prior beliefs
217 that the probabilities of desirable and undesirable events occurring under their control
218 were low. Contrast this with non-depressed participants that may have come in with
219 a belief that there was a higher probability of desirable events occurring under their
220 control compared with undesirable events. From a Bayesian perspective, the different
221 attributions of control by depressed and non-depressed participants are thus rationally
222 motivated. They reflect prior beliefs that have received support over the lifetime of
223 experiences, and inform judgments of the controllability of novel situations.

224 As stressed above, further research is required to precisely determine the role of
225 priors. Given, however, the potential for the Bayesian approach to provide a uni-
226 fying framework for existing demonstrations of the IoC, we believe there are great
227 gains associated with such an empirical program. It is clear to us that a conflation
228 of skill and chance based situations can result in an increased likelihood of people
229 demonstrating an IoC (e.g., [Langer 1975](#)). Likewise, in repeated choice tasks an asso-
230 ciative learning account (e.g., [Matute et al. 2007](#)) provides a good account of the
231 data, and could well be the best explanation for the data in such contexts. Moreover,
232 associations between active involvement and frequent positively reinforced outcomes
233 can be explained using a control heuristic in which people's judgments of control

234 are informed by the perceived connection between their own action and the desired
235 outcome, and their intention to achieve the outcome (Thompson et al. 1998). Rather
236 than a critique of existing accounts, our intention has been to demonstrate that no extant
237 account is able to explain all manifestations of the IoC observed in the literature. The
238 Bayesian framework would appear to be a suitable candidate for providing such an
239 overarching framework, and furthermore would link IoC research to a broad literature
240 investigating the probabilistic nature of cognition (see e.g., Chater & Oaksford, 2008;
241 Chater et al. 2006, 2010; Griffiths et al. 2010).

242 4 Combining priors and evidence: an empirical example

243 Ladouceur and Sevigny (2005) investigated the effect of a ‘stopping device’ on the IoC
244 in a video lottery task divided into two phases. In Phase 1, participants were simply
245 told to hit the ‘play’ button and then to watch as the reels settled on certain symbols.
246 When asked, no participants reported that it was ‘possible to influence the symbols
247 after having pressed play’, that ‘there was a method for controlling the outcome’, that
248 ‘skill could contribute to winning’, or that ‘there were strategies that could enable one
249 to increase winning chances’. In Phase 2, participants were told that, if they desired,
250 they could stop the reels by pressing anywhere on the computer screen. In this phase
251 participants were more likely to answer the above questions in the affirmative, and
252 in a follow up experiment participants who were given a stopping device played the
253 game significantly longer than participants not provided with a stopping device.

254 The responses of these participants appear entirely rational. Under conditions in
255 which there are no obvious cues indicating control, participants correctly report that
256 there is no potential to control the outcome, whereas when there are obvious cues
257 (i.e. a stopping device), participants give answers that are perfectly consistent with
258 (erroneously) perceiving that it is *possible* to influence the outcome, and consistent
259 with this, they behave accordingly, by playing for longer. Participants might know
260 that they are, as of yet, unable to influence the outcome, but the evidence from the
261 task suggests that it is controllable in principle. Perhaps if they practice for longer
262 they can achieve control over the outcome. After all, many everyday tasks might at
263 first appear uncontrollable, but we subsequently learn to control them (control over
264 our own bladder is one of the earliest examples). In this study, participants used the
265 evidence in the task and their past experience with the need to ‘learn’ controllability
266 to infer that a currently uncontrollable task might, nevertheless, be controllable in the
267 future.

268 5 Controllable or uncontrollable?

269 Bayes’ Theorem alone cannot determine those situations in which one should act as
270 though a scenario is controllable. Whether or not to act in such a way is a binary
271 choice, whilst Bayes’ Theorem’s output is a probabilistic degree of belief. A binary
272 output from a ratio scale requires the setting of some threshold value. Decision
273 Theory prescribes that this output reflects both degrees of belief (probabilities) and
274 the utility (subjective goodness or badness) of different possible outcomes. In any

Table 1 The four possible outcomes of decisions based upon the controllability of a situation

		State of the world controllable	Uncontrollable
Act as though the world is...	Controllable	<i>Hit</i>	<i>False positive (Illusion of Control)</i>
	Uncontrollable	<i>Miss (Illusion of Chaos)</i>	<i>Correct rejection</i>

275 situation, there are two possible states of the world: The situation is controllable or the
 276 situation is not controllable. Similarly, one can act as though the world is controllable
 277 or uncontrollable. Thus, there are four potential outcomes (see Table 1).

278 Correct classifications of the world as controllable or uncontrollable (hits and cor-
 279 rect rejections) are associated with positive utility, whilst incorrect classifications of
 280 the world (false positives and misses) are associated with negative utility. If the posi-
 281 tive utility associated with a hit equals the positive utility associated with a correct
 282 rejection and the negative utilities associated with false positives and misses match,
 283 then the threshold for deciding that a situation is controllable or uncontrollable is:
 284 $P(\text{controllable}) = .5$. As soon as there is an asymmetry in the utilities, however,
 285 then the threshold moves from .5. Thus, if one perceives higher costs with a miss than
 286 with a false positive, the threshold will be less than .5, increasing the likelihood of
 287 acting as though a situation is controllable.

288 6 The London riots

289 Between August 6th and August 8th, 2011, London experienced its worst rioting for
 290 over thirty years. Rioters felt untouchable, and law-abiding citizens lamented that the
 291 situation appeared out of the control of police or government. Given scenes of grossly
 292 outnumbered police forced to stand by and watch the rioters, along with the lack of
 293 arrests at the time, and the sheer scale of the criminality, the evidence available pointed
 294 to an uncontrollable situation. On August 9th, the British Prime Minister, David
 295 Cameron, stated "...people should be in no doubt that we will do everything neces-
 296 sary to restore order to Britain's streets and to make them safe for the law-abiding..."
 297 Despite the available evidence at the time, this situation *did* prove to be controllable,
 298 and police were able to contain the situation. This example shows the benefits of per-
 299 ceiving a potential for control over a situation (sometimes in the absence of strong
 300 evidence for controllability).

301 If viewed in decision theoretic terms, there is a huge cost attached to signaling to
 302 the UK population that the events surrounding the riots provide sufficient evidence of
 303 an uncontrollable situation. Therefore, if wrongly judged to be uncontrollable, then
 304 no additional preventive action will be taken, and so further rioting and chaos ensues.
 305 In other words, it seems as though the costs associated with perceiving a controlla-
 306 ble situation to be uncontrollable (illusion of chaos) will often be greater than those
 307 associated with perceiving an uncontrollable situation to be controllable (illusion of
 308 control). Consequently, in addition to often being in possession of priors and evidence
 309 indicative of a controllable situation, it might also be considered rational to set the

310 threshold for acting as though a situation is controllable at less than .5, further exag-
311 gerating the likelihood of erroneously, but rationally, classifying an uncontrollable
312 situation as controllable (as in IoC tasks).

313 Scheibehenne et al. (2011) highlighted a similar asymmetry related to the expect-
314 ation for positive recency (the “Hot Hand Fallacy”) in prediction tasks. Expecting or
315 perceiving a pattern erroneously in such tasks cannot harm one, as performance cannot
316 be better than chance. However, not spotting a correct pattern prevents an agent from
317 being able to capitalise on the predictiveness of the environment.

318 Matute and colleagues (e.g., Matute 1996; Matute et al. 2007) make a similar point
319 to the one advanced in this section, recognising that to obtain a desired outcome it is
320 likely to be beneficial to continue to act (rather than be passive, and see if you obtain
321 the outcome anyway). However, on Matute and colleagues’ account, this is seen as a
322 precursor to the IoC, which then occurs because participants act continuously to obtain
323 the outcome and thus cannot learn that the outcome would also have occurred in the
324 absence of their action. On our account, participants alter their threshold for deciding
325 that a situation should be classified as controllable or uncontrollable, having already
326 made a judgment of the likely controllability of the situation.

327 The present account of the IoC also exists within a broader framework outlined in
328 McKay and Dennett (2009). They claim that there is a functional role in acting on
329 misbeliefs that are based on reliable generalizations (e.g. men’s general perception
330 of women’s sexual interests in them) that can be anomalous in particular situations
331 (e.g. a woman doesn’t share this intent). The trade-off between the overall accuracy of
332 beliefs and their inaccuracy in certain situations comes from an asymmetry between
333 the benefits (i.e. fitness) and costs (i.e. decreased fitness) of actions. Similarly, the
334 IoC is a candidate for what they would classify as a (mis)belief that is an adaptive
335 by-product of a cognitive system.

336 7 Conclusion

337 The British Prime Minister’s announcement at the time of the riots illustrates the need
338 to be seen as being ‘in control.’ Moreover, it is an illustration of the profound impact
339 that perceived control has in affecting real world outcomes. We have argued here that
340 the IoC is not a product of biased, motivated reasoning, but can be seen as a product
341 of a rational Bayesian reasoning process. By reviewing the IoC literature we argue
342 that forming judgments of control generally involves marshaling relevant evidence,
343 which is combined via Bayesian methods of inference with prior beliefs that people
344 bring to laboratory settings. Furthermore, in making a binary judgment as to whether
345 to act as though a situation is controllable or not, people operate under the assump-
346 tion that overestimating control in uncontrollable environments is advantageous to
347 underestimating control in controllable environments.

348 References

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uncorrected proof