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2 On the reliability of unreliable information: Gossip as cultural
3 memory

4
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7 **Abstract**

8 When individuals learn from what others tell them, the information is subject to transmission
9 error that does not arise in learning from direct experience. Yet evidence shows that humans
10 consistently prefer this apparently more unreliable source of information. We examine the effect
11 this preference has in cases where the information concerns a judgment on others' behaviour and is
12 used to establish cooperation in a society. We present a spatial model confirming that cooperation
13 can be sustained by gossip containing a high degree of uncertainty. Accuracy alone does not
14 predict the value of information in evolutionary terms; *relevance*, the impact of information on
15 behavioural outcomes, must also be considered. We then show that once relevance is incorporated
16 as a criterion, second-hand information can no longer be discounted on the basis of its poor fidelity
17 alone. Finally we show that the relative importance of accuracy and relevance depends on factors
18 of life history and demography.

19 **Keywords:** gossip, reputation, cooperation, social norm

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1 Introduction

A cooperative society is defined as one in which individuals benefit from the collective absence of defection (Axelrod and Hamilton, 1981). For any individual member defection is advantageous only if it is the exception. An example of cooperation is free trade between nations (Schelling, 1960). Against a context of free trade any one nation will benefit by introducing import tariffs but if all do so, all lose (Hardin, 1968). This paradox makes cooperation vulnerable, thus the discovery of cooperation in many different animal societies is surprising. Evolution depends on the relative fitness of individuals so we might expect individuals to maximize this regardless of its consequences for the group (Williams, 1966; Dawkins, 1976). The existence of cooperation under such conditions can be explained by direct reciprocity (Trivers, 1971): if interactions in the society are between those who have met before and will meet again, a you-scratch-my-back rule between pairs results in a society which meets the criteria for cooperation. However cooperation also exists in societies such as contemporary urban ones, where populations are large and mobile and interactions are between strangers that are unlikely to meet again. Here direct reciprocity fails as an explanation, as the society is vulnerable to free-riders—individuals who defect opportunistically and move on before the consequences of their behaviour can catch up with them (Enquist and Leimar, 1993). Although such lifestyles of course have their own costs and benefits (Taylor, 2014), in such cases a different mechanism is typically used to explain cooperation. *Indirect reciprocity* (Alexander, 1987; Nowak and Sigmund, 1998, 2005) is a system in which agents do good to one other because they are likely to receive good from a different other. One way this can be achieved is when individuals observe each other’s behaviour and pass on the resulting reputation in social transmission—gossip. Defectors can no longer free ride; however mobile they are they risk being preceded in their interactions by their reputation (Milinski et al., 2006).

In relationships based on direct reciprocation we can imagine that it is quite possible for an individual to form a judgment of the quality of a partner because the history of previous interactions provides its own guarantee. But in order for a system of indirect reciprocity to hold, a similar quality of information is required of partners who have never met before (McElreath and Boyd, 2007). Requiring a reputation to achieve indirect reciprocity places greater demands on it (Roberts, 2008). Whereas direct reciprocity requires memory of previous experience of direct perception, here the information is a temporally and spatially displaced judgment determined and passed on by others and as such potentially more vulnerable to error. In this article, we focus on the question of how gossip can support cooperation (Nowak and Sigmund, 2005) despite this potential for inaccuracy.

51 In cases where the reputation of a partner is learnt third-hand via social transmission of information
52 (gossip) the reliability of that information is paramount. We might expect individuals to be very
53 sensitive to any evidence of error in the information. However a number of researchers have shown
54 that our behaviour contradicts this expectation. Human societies continue to be trusting of socially
55 learned information even when it is known to contain much error and uncertainty (Richerson and
56 Boyd, 2005; Hess and Hagen, 2006). What is more, when faced with the direct evidence of our senses
57 contradicting social learning we often give preference to the latter (Asch, 1956) even when based on
58 anonymous gossip (Sommerfeld et al., 2007). Comparisons of human and non-human cognitive abilities
59 (Nagell et al., 1993; Whiten et al., 2009) cite humans' greater use of culture—which is founded on social
60 learning—as our key advantage (Gergely and Csibra, 2006; Henrich and McElreath, 2003; Tomasello,
61 1999). Human infants when taught to perform a task in an inefficient manner persist in it despite
62 exposure to a better alternative (Gergely et al., 2002; Meltzoff, 1988), while other primates imitate
63 only necessary actions once some demonstrated actions are recognised as unnecessary (Whiten et al.,
64 2009). In this article we present a possible explanation for the human propensity to over-imitate, by
65 analysing the information in gossip relative to information from direct experience.

66 Humans intuitively value accuracy in information and have a variety of methods to assess it. We
67 assign lower value to information that is contradictory, or that differs from an authoritative source.
68 Scott-Phillips (2008) argues that in nature information should be defined in evolutionary terms. But
69 evolution does not access the accuracy of information directly, it acts on the results of the information
70 being put to use (Rauwolf et al., 2015). This has the effect of introducing *relevance* (Dessalles, 2007;
71 Sperber and Wilson, 1995) as a crucial component of information value. We define relevant information
72 as information that is needed to perform the task at hand. Information that is highly accurate but
73 unlikely to be put to use will therefore have less value. Non-relevant information will not be acted
74 upon, and therefore will not be subject to selection pressures, whether it is accurate or inaccurate.

75 When individuals rely on experience, we assume accuracy is not in question but the relevance
76 of what they learn is dictated by the limits of the life history of the individual concerned (Čač
77 and Bryson, 2007; Roberts, 2008). In the case we examine—the role of reputational information in
78 successive interactions—the information in question can only be put to use when an individual meets
79 the same individual again and the consequences of reputation gained earlier are realized. The notion

80 of relevance in this context is captured by the cost / benefit relationship in the following equation:

$$c < wb \tag{1}$$

81 where w is the probability of future interactions, c is the cost incurred at the first interaction, and b
82 is the benefit expected from that cost in any future interactions (Roberts, 2008). When an individual
83 learns by social communication, relevance is likely to be higher as social transmission decouples the
84 information from full dependence on physical experience. In the limit case of instant communication,
85 $w = 1$, agents can be sure to know the most recent social actions of any individual they meet. This
86 does not mean that social information is always of relevance to the audience of gossip, but that
87 on the occasions when an interaction is influenced by the reputation of an agent there is a higher
88 probability of benefit from information if it is socially transmitted than if it must be learnt through
89 individual experience. This is because it is more likely that others may have had recent (or indeed
90 any) experience of the interactor. However in freeing information about interactions from spatial and
91 temporal constraints, socially transmitted information may incur a cost due to loss of accuracy. We
92 are agnostic here as to whether error is introduced deliberately or by misperception; we only assume
93 that error accumulates over the process of repeated transmission. We can again describe a necessary
94 relationship:

$$q < wb/c \tag{2}$$

95 where q is the probability of the reputational information being correct. That is, the expected future
96 benefit must outweigh the cost of the interaction by a sufficient extent to compensate for the probability
97 of miscommunication.

98 In a series of three experiments we explore the effect of incorporating both relevance and accu-
99 racy in the evaluation of information learned by direct experience or by social communication. First
100 we replicate in a spatial simulation the mathematic result of Nakamura and Masuda (2011). This
101 demonstrates that in the full social communication case ($w = 1$) reputational information can sustain
102 cooperation despite containing a large degree of uncertainty caused by assigning others reputations
103 based on incomplete observation. In Experiment 2 we relax the ($w = 1$) assumption by parameteriz-
104 ing the rate of spread of social information, and substituting low transmission fidelity for incomplete
105 observation as the cause of error. We show that accuracy alone cannot predict the ability of social
106 communication to sustain cooperation. A level of accuracy too low to sustain cooperation at a lower

107 rate of transmission is nevertheless able to do so if the transmission rate is increased. In the third
108 experiment we replace control of transmission rate with the assumption of a limited lifespan for agents.
109 We compare the performance of individuals that learn from direct experience with those that use so-
110 cial communication. A limited lifespan means that reputational information is liable to be outdated
111 and thus of lower relevance. But as the direct experience strategy can be more vulnerable in this
112 respect we again show that accuracy alone cannot predict the outcome. A gossiping strategy which is
113 ruled out by its high level of error under conditions of long or unlimited lifespan may still outperform
114 experience-based learning if lifespans are shorter. Our conclusion is that the relative merits of rele-
115 vance and accuracy are highly sensitive to the ecology of the society. Adhering to a gossiping strategy
116 despite the potential inaccuracy of the information can thus be evolutionarily successful under many
117 conditions.

118 **2 Methods**

119 **Model**

120 All experiments are based on the same agent-based model of evolutionary game theory. Interaction
121 between individuals following different strategies and influenced by different kinds of information gen-
122 erates varying payoffs which are used to evolve the population.

123 **Individual Interactions: the Donation Game**

124 Following the method used by Nakamura and Masuda (2011), agents in our simulation play the dona-
125 tion game, which is similar in form to the prisoner’s dilemma (See Figure 1). The idea of the game
126 is based on ordinary social interactions, where an individual can choose to help a passing stranger
127 or not, for example by giving directions. Individuals move at random in a two-dimensional world at
128 each iteration. The world is a 21 x 21 torus which wraps horizontally and vertically. If two agents are
129 located adjacently, roles for the game are assigned randomly and they are able to interact by playing
130 a move in the game. The game has two players: a donor and a recipient. There are two possible
131 moves in the game. If the donor cooperates, it pays cost c and the recipient gains benefit b . If the
132 donor defects it pays nothing and the recipient gains nothing. In cases where $0 < c < b$ the dilemma
133 of cooperation described in the introduction is created: in a single interaction the best move for the

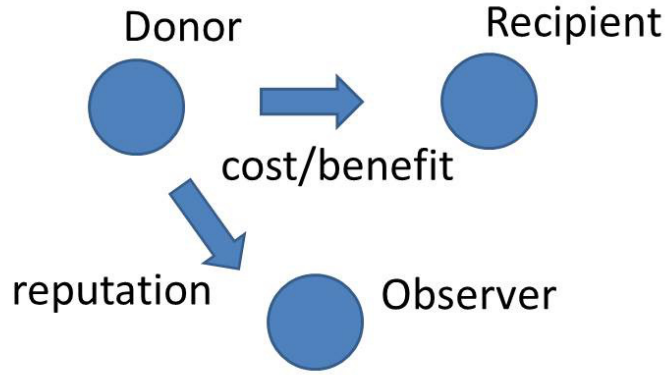


Figure 1: The donation game involves three actors: a *donor* who chooses whether or not to donate to a *recipient*, and an *observer*. Where a donation occurs, the donor pays c and the recipient receives b . The observer updates its own record of the donor’s reputation both on the basis of the donor’s action, and possibly on the reputation the observer holds for the recipient (see main text and Table 1.)

Table 1: Social norm used in experiments 2 and 3. The second-order judging norm.

	Good	Bad
Cooperate	G	B
Defect	B	G

134 donor is to defect, but each individual does better if all cooperate.

135 Reputation Assigned by Observers According to a Social Norm

136 Individuals have binary reputation values: Good or Bad. The observer updates the donor’s reputation
 137 based on a combination of the donor’s action: Cooperate or Defect, and the recipient’s reputation
 138 according to a single model-wide rule called the social norm. Here we use the judging norm (Nakamura
 139 and Masuda, 2011; Ohtsuki and Iwasa, 2006, see Table 1). This specifies a society which approves of
 140 those who cooperate (C) with the good (G) and defect (D) against the bad (B). Each individual has
 141 its own record of others’ reputations. Individuals meet at random and update each other’s records so
 142 that reputation spreads. For a discussion of other possible norms and their effects on donation games,

143 see Rauwolf et al. (2015).

144 **Reputation Impacts on Individual Choice of Action**

145 The donors' choice of action is determined by their action rule (Nakamura and Masuda, 2011). There
146 are 3 populations each acting according to a different rule: unconditional cooperators (ALLC agents)
147 cooperate regardless of their partner's reputation; unconditional defectors (ALLD agents) defect regard-
148 less of their partner's reputation, and discriminators (DISC agents) cooperate with partners who are
149 good but defect against bad partners (Ohtsuki et al., 2009).

150 **Evolution According to Payoff**

151 The population is reproduced according to fitness-proportionate selection (Goldberg and Deb, 1991)
152 based on pay-offs from the donation game. Mutation, in which an action rule different to the parent's
153 is substituted in offspring, occurs according to a probability (see Table 2).

154 **Experiment 1. Judgments Made Without Full Knowledge**

155 It can be seen from Table 1 that the observer's assignment of reputation to a donor is a second-order
156 judgement. It is determined by a combination of the donor's action with the recipient's reputation.
157 Nakamura and Masuda (2011) introduce uncertainty in the second term, so that a proportion of donors
158 take part in and are judged on interactions with recipients of unknown reputation. The authors
159 identify three social norms which are still able to realize indirect reciprocity under this condition. We
160 replicate this in the case of a single norm—the suspicious judging norm—since rather than considering
161 the relative performance of different norms we are interested in the extent to which a given norm
162 can continue to function under conditions of unreliable information. We choose the suspicious norm
163 motivated by the finding in Yoo (2009) that negative gossip has a greater impact on the listener.

164 **Method**

165 The donor and observer only know the reputation of the recipient with probability $0 < k < 1$. This
166 is concomitant: all agents share the same information, so either they both know or neither do. In
167 view of this the action rules are extended. ALLC agents here also cooperate (C) with partners whose
168 actual reputation is unknown (U) to them. ALLD agents defect (D) against unknown partners. DISC

169 agents cooperate with unknown (U) partners. The social norm is also extended (see Table 2). Players
170 who cooperate with unknown partners are judged good (G) but those who defect against an unknown
171 partner are judged bad (B). This norm is termed suspicious because faced with an agent who defects
172 against a partner unknown to him observers have a choice of two assumptions. The trusting assumption
173 is that he defects because the unknown agent is in fact bad. The suspicious one is that he defects
174 despite the fact that his partner is in reality good.

175 **Simulation**

176 In each round, two randomly selected agents are assigned the roles donor and recipient and play the
177 donation game. Payoffs to both players are determined according to the rules of the game described
178 above. To begin by considering the most extreme case (this will be refined below), reputation is spread
179 instantly and with perfect transmission fidelity to all agents resulting in a single global reputation for
180 each agent which is then used to play the next round of donation game. An assumed observer updates
181 the donor's reputation in the list held by all agents according to the social norm described above. The
182 list records each agent as either good or bad¹. Rounds are repeated until every agent has played once
183 against every other. We call this one generation. At the end of each generation the sum of all the
184 costs and benefits to each agent are calculated.

185 A generation is followed by a round of reproduction. The strategy of each agent is encoded in a
186 genome of one position with three alleles. Each member of the new generation is selected from the
187 previous generation according to fitness-proportionate selection. The chance of selecting individual i
188 is $p_i / (\sum_{j=1}^n p_j)$, that is the payoff of individual i divided by the total pay-off of the population. There
189 is also a small chance for mutation $u = 0.01$, where an agent's strategy is replaced by one of the three
190 available strategies. At the start of the run the population is made up entirely of DISC agents. This
191 process iterates 2000 times to provide ALLD and ALLC sufficient opportunity to invade DISC from
192 scarcity.

193

194

¹In experiments 2 and 3, global reputation is replaced by peer-to-peer gossip and individual memory in order provide a setting in which claims made about cheap communication can be examined.

Table 2: The suspicious social norm, which is used in Experiment 1. The judging norm is extended to include a first-order norm in cases where the recipient’s reputation is unknown. The norm assumes that when a donor plays Defect against a partner with unknown reputation he is playing Defect to a Good partner, hence is suspicious of him, rather than the alternative possibility that he is playing Defect because his partner’s reputation is Bad.

	Good	Bad	Unknown
Cooperate	<i>G</i>	<i>B</i>	<i>G</i>
Defect	<i>B</i>	<i>G</i>	<i>B</i>

Table 3: Table of free parameters, values used in present figures, and a sensitivity report including range of values tested.

Sym- bol	Parameter	Value	Notes
n	population size	100	number of agents.
r	rounds	10,000	donation games per generation.
g	generations	5,000	number of evolutionary iterations.
c	cost	1	cost of cooperating, kept constant.
b	benefit	[1,10]	benefit of being recipient of cooperating donor, increments of 0.25
<i>u</i>	mutation rate	0.01	chance of strategy mutation.
k	knowledge	[0,1]	probability recipient’s reputation known, increments of 0.025

195 Results and Discussion

196 In the unshaded parameter region in Figure 2 interactions are entirely cooperative. A population which
 197 is predominantly DISC in the context of these parameter ranges and the suspicious judging social norm
 198 produces a society where interactions are overwhelmingly cooperative (C). The result is consistent with
 199 that of Nakamura and Masuda (2011). Cooperation is stable against invasion down to low values of k
 200 provided that the cost/benefit ratio is high. Conversely in regions where the cost/benefit ratio is low
 201 a higher value of k is required to enable cooperation.

202 These results suggest that within a specified context of action rules and social norms cooperation can
 203 persist in a society even when a significant amount of the information used to enforce the cooperation is
 204 uncertain. The assumption of trust specifies a cooperative action from the donor when the recipient’s
 205 reputation is unknown (Nakamura and Masuda, 2011). This means that a certain proportion of
 206 recipients whose actual reputation is Bad will escape the punishment. That cooperators are able to
 207 overcome such limitations on observation demonstrates that it can be in an individual’s interest both
 208 to tolerate being judged and to judge others, even in this imperfect fashion.

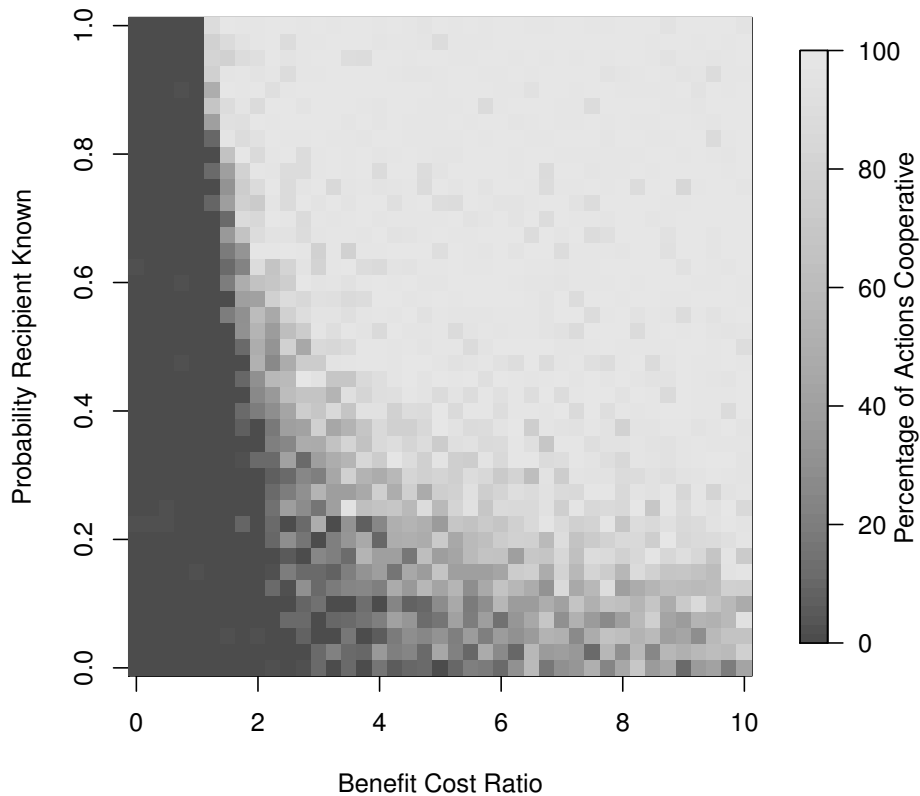


Figure 2: Judgments made without full knowledge. Percentage of cooperative interactions aggregated over 10 runs. In the unshaded parameter region interactions are cooperative. Dark cells indicate that at 2000 generations DISC is invaded by ALLD.

b

209 **Experiment 2. Agents' Knowledge of Reputations Differs Ac-**
 210 **cording to the Spread of Gossip**

211 In Experiment 1 the information carried by gossip was sometimes incorrect at the source but there was
 212 no error in gossip transmission. But we know that in the real world, knowledge and behaviour take
 213 time to diffuse through a population (Gruber et al., 2015). Here we investigate error in transmission
 214 and the speed at which information is spread. We replace the idealized broadcast transmission of Naka-
 215 mura and Masuda (2011) with a more ecologically plausible spatial model of peer-to-peer interactions
 216 occurring during encounters between mobile agents.

217 Method

218 Agents interact when they are in sufficient proximity to each other. Specifically the world is an
219 21x21 grid, and agents interact on each cycle with at most one other agent, if one is present on an
220 adjacent grid cell. After playing a round of the donation game the partners exchange information.
221 This communication takes place regardless of the nature of the interaction that has just taken place.
222 If one agent believes a third party to be bad, and the other believes it to be good, then the agent with
223 the ‘good’ association updates this instead to ‘bad’. We use this pessimistic rule to represent a society
224 in which reputation is difficult to achieve but easy to lose.

225 In Experiment 2, the explicit probability used in Experiment 1 ($1 - k$) that a recipient in an
226 observed game may be unknown (U) is replaced by the emergent chance that information about a
227 player’s reputation may not have spread sufficiently fast to inform the donor or observer in their next
228 interaction. All agents are initially assigned a good reputation, we examine how efficiently evidence
229 of infringed reputation is spread. Therefore, agents no longer require individual behaviour rules for
230 handling unknown (U) individuals. The probability of information being up-to-date is a function of
231 the agents’ collective behaviour. The rate at which gossip spreads is a function of a limit t placed on
232 the number of individuals whose reputations may be exchanged at each encounter, see Table 2. This
233 limit combined with the absence of a global reputation means that agents do not know each other’s
234 knowledge state and therefore cannot selectively divulge full information. Rather, agents communicate
235 a random set of their own knowledge. Note that it is possible for the gossip to have reached the donor
236 but not the observer or vice versa. This type of observation is known as *independent*, in contrast to
237 Experiment 1 where due to the use of broadcast reputation observation was concomitant.

238 An error rate e specifies the probability that at each exchange of gossip the opposite update to the
239 one intended is made by the recipient (see Table 2.) Populations of the three strategies (ALLC, ALLD,
240 DISC) are of equal size throughout Experiment 2. There is no reproduction, selection or mutation.
241 The relative performance of strategies is judged on their average pay-off from the donation game

242

243 Results and Discussion

244 Figure 3 shows that minimum values of both transmission rate and accuracy are required in order for
245 DISC to out-perform ALLC and ALLD. However within this context there is an interplay between the

Table 4: Table of free parameters, values used in figure 3, including range of values tested.

Sym- bol	Parameter	Value	Notes
n	population size	150	number of agents.
r	rounds	1,000	spatial model so number of interactions per round varies
c	cost	1	cost of cooperating, kept constant.
b	benefit	3	benefit from being recipient of cooperating donor, kept constant
e	error	[0, 100]	per cent chance incorrect transmission of reputation
t	transmission	[0, 100]	percentage of population gossiped about at each encounter.

246 two parameters. A level of accuracy poor enough to cause DISC to fail at low transmission rates is
 247 nevertheless sufficient to make the DISC the winning strategy once transmission rate is raised.

248 When accuracy and relevance are both factors in the value of information, socially-learned infor-
 249 mation even with a high incidence of error may outperform experience-based learning if the latter
 250 is insufficiently relevant (Sommerfeld et al., 2007). By parameterizing the rate at which information
 251 spreads, we are able to examine the interaction between relevance and accuracy. Although relevance is
 252 not directly controlled in our agent-based model, as transmission rate increases so does the probability
 253 of having reputational information concerning an individual that an agent is interacting with. The
 254 risk of being updated by inaccurate information is outweighed by the risk that not being updated will
 255 cause greater inaccuracy.

256 **Formal analysis of an ALLD agent’s reputation**

257 To better understand how transmission and error rate affect the performance of each strategy, we turn
 258 to formal analysis. Cooperation is maintained by eliminating free-riders, and this is only accomplished
 259 if free-riders are negatively reputed. Here we analyse the probability, $ALLD_b$, that an ALLD is
 260 considered bad.

261 To do this formally, we presume a few simplifications. Gossip and the spread of social information
 262 inherently rely on the specifics of a social network. If an individual sees one person more than another,
 263 they are more likely to receive information from that person. While computer simulations easily permit
 264 spatial and network dependencies, here we focus on elucidating the interplay between transmission and
 265 error rate. In doing this, the nuances of a particular social network are abstracted away. However, as
 266 a consequence we slightly alter the definition of error rate and transmission rate.

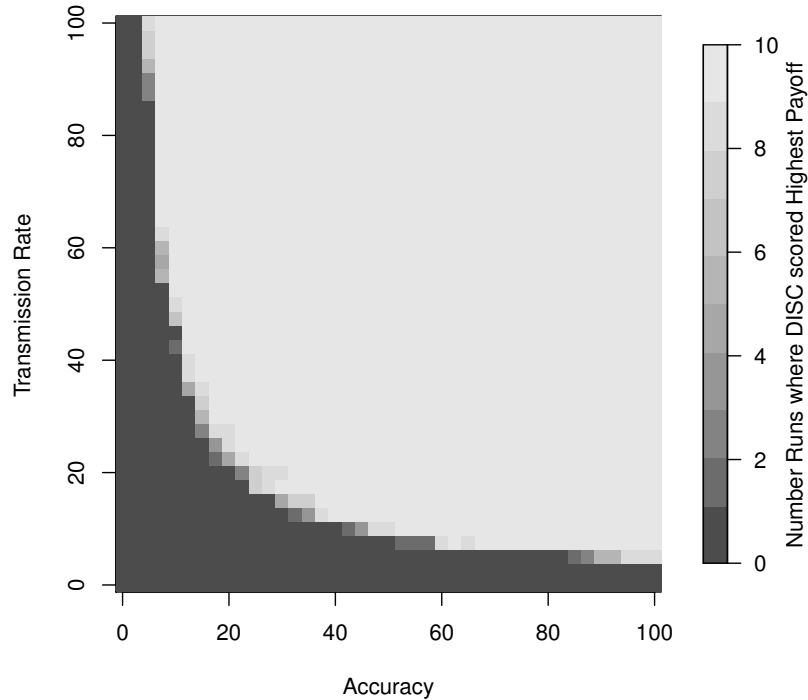


Figure 3: Experiment 2, in which agents’ knowledge of reputation differs as a result of the spread of knowledge. Simulation results are shown for a range of error rates in transmission (e), and rates of transmission (t), aggregated over 10 runs. In the light region DISC scores a higher payoff than both ALLC and ALLD.

267 When an agent uncovers a free-rider, it attempts to pass on this information through gossip. We
 268 define transmission rate, t' , as the probability an agent will hear the newly discovered information
 269 during a round of play. Previously, transmission rate was defined as the chance that two interacting
 270 agents would gossip about another agent. The gossip could then be passed to other agents at some later
 271 interaction. However, the fact that these simulations are spatial—that is, each agent has a location and
 272 this determines which agents they interact with—has consequences. While transmission of gossip relies
 273 on whether or not a bit of information is discussed, it also relies on the probability of interacting with
 274 other agents with relevant knowledge. This is correlated with location. Here we define transmission
 275 rate to encapsulate both space and the chance to spread gossip. t' is the chance that, in any round,
 276 an agent will hear a specific piece of gossip.

277 When an agent hears gossip there is a chance it will be in error. e' , is the chance communication
 278 is erroneous. Again, this is slightly different compared to the error rate employed in the computer

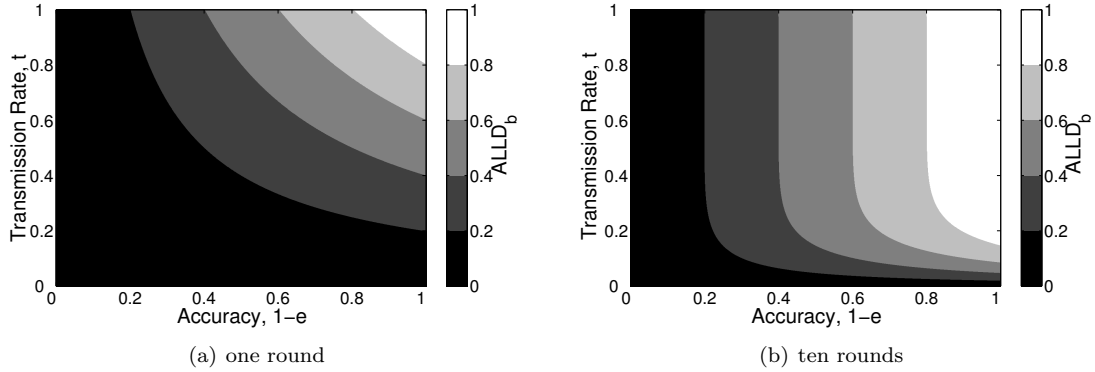


Figure 4: The probability $ALLD_b$ that an agent has heard that a particular ALLD agent is bad, after (a) one and (b) ten rounds of play. Mathematical outcomes are shown for a range of error rates in transmission (e') and rates of transmission (t').

279 simulations. Previously, error rate represented the chance that at a given communication was incorrect.
 280 Here we presume that whenever an agent hears information, there is an $1 - e'$ chance that it is correct.
 281 If gossip travelled instantaneously, then $1 - e'$ fraction of agents would correctly diagnose the defec-
 282 tor. However, information does not spread instantaneously, it depends on the speed of transmission.
 283 So, if an agent discovers a defector, the chance this information will accurately reach another agent in
 284 a given round is:

$$ALLD_b = (1 - e')t' \quad (3)$$

285 An ALLD is marked as bad if the agent receives the information, t' , and the information is correct,
 286 $1 - e'$. Furthermore, the probability that an agent will accurately hear the gossip after i rounds can
 287 be written as:

$$ALLD_b = (1 - e')(1 - (1 - t')^i) \quad (4)$$

288 There is a $1 - t'$ chance that an agent does not hear the information in a given round. Consequently,
 289 there is a $(1 - t')^i$ chance that an agent has not heard the information after i rounds. $1 - (1 - t')^i$ is
 290 then the probability that an agent has heard the information in i rounds. This result is then multiplied
 291 by the probability that the information is accurate, $1 - e'$.

Results and Discussion

A parameter sweep of e' $[0, 1]$ and t' $[0, 1]$ can be plugged into Equation 4 to calculate the likelihood than an ALLD agent is given a bad reputation, $ALLD_b$. Figure 4(a) shows $ALLD_b$ after one round of gossip. Clearly, $ALLD_b$ depends on the interplay between the transmission and error rate. As the transmission rate increases, so does the likelihood of speedily informing others. However, the accuracy of the transmission is limited by the error in communication.

Figure 4(b) shows the percentage of negatively reputed ALLD agents after 10 rounds of gossip. As time passes, the effects of the transmission rate diminish. After ten rounds, an agent inhabiting a relatively low-error-rate world is likely to have heard the information, regardless the transmission rate. Only very small levels of t' show a marked difference.

However, whilst the effect of the transmission rate may become moot after several rounds, cooperation might not last that long. If an ALLD is permitted to operate with a good reputation for multiple rounds, then it might undermine a cooperative society long before future rounds can be played. This is why, in Figure 2, the strategy with the highest pay-off depends on the rate of transmission. Expedited and relevant information dissemination reduces the opportunity for free-riding, increasing the pay-offs of DISC agents.

Experiment 3. Personal Experience Versus Gossip Under Limited Lifespan

In Experiment 2 the value of information about reputations did not degrade over time. The information an individual has may be incorrect either because of transmission error or because restriction on the transmission rate prevents incorrect information from being updated. But there is another way in which information could be false which was not considered: that is, in the time since the information originated the object of its reference may have changed. Here we revoke this assumption and give the value of the reputational information a time limit. Information that was originally relevant may subsequently become irrelevant when the state of affairs it applies to has changed. We do this by imposing a limited lifespan on the agents, affecting both the individuals that carry information and the individuals that the information refers to.

Table 5: Table of free parameters, values used in Figures 5 and 6, and values tested.

Sym- bol	Parameter	Value	Notes
n	population size	200	number of agents.
r	rounds	5,000	spatial model so number of interactions per round varies
e	error	[0.7]	per cent chance incorrect transmission of reputation
l	lifespan	[3000, 10000]	at initiation each individual's life set at random random between 0 and this upper limit

320 Method

321 In this simulation we compare two strategies: *Direct Observers* and *Gossipers*. Direct Observers do
 322 not transmit gossip and do not receive it. Their only source of information is their own observations.
 323 Here we assume that observer's information on others' reputations is perfectly accurate, being based
 324 on their own (perfect) perception, and not subject to transmission error e . This assumption is the limit
 325 case, giving the observers the maximum advantage of their strategy. In contrast, Gossipers transmit
 326 and receive gossip about others' reputations. The accuracy of their information is affected by the value
 327 of e and the number of gossiping interactions that the information has crossed.

328 We introduce a lifespan value l that determines the maximum number of rounds an agent lives.
 329 Whether lifespan is a genetically regulated trait is a matter of some controversy, lifespan is a common
 330 feature used to characterise species, and from a theoretical perspective it has known impacts (Kokko,
 331 1997; Čače and Bryson, 2007). At the beginning of a simulation, each agent is randomly assigned an
 332 age within the lifespan. After each round the lifespan reduces by 1. When an agent reaches the end of
 333 its life it is deleted. Dead agents are replaced by new agents which possess no information from either
 334 gossip or observation. Other agents in turn have no information from direct observation or gossip
 335 regarding the reputation of new agents.

336 As in Experiment 2 at the start of the run agents' reputations are set to good (G) and gossip consists
 337 of setting good (G) to bad (B) only. This provides a simple rule for which agent takes precedence in
 338 information exchanges. As in Experiment 2, populations of all six strategies are of equal size. Note
 339 that as in Experiment 2 there is no selection or mutation, even though there *is* reproduction. New
 340 agents are identical clones of their parent to limit the experiment to testing the effect of irrelevance
 341 generated by agent turnover. The relative performance of Direct Observation versus Gossip is judged
 342 on the extent to which each population has assigned bad reputations (B) to ALLD (defector) agents.

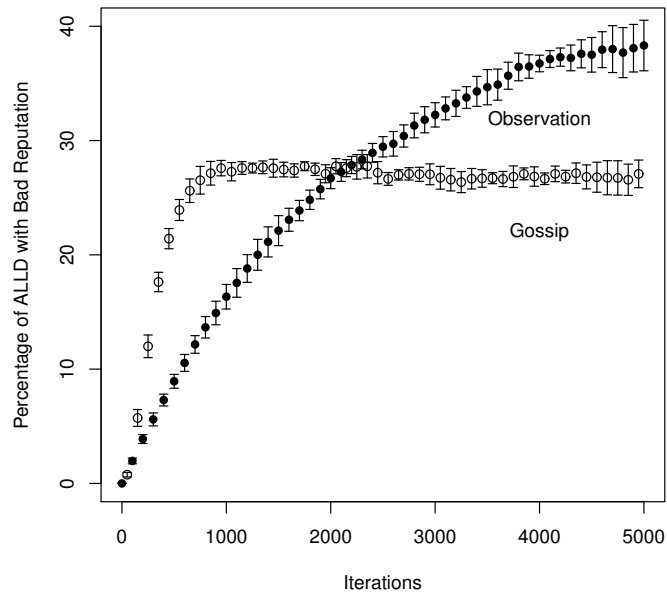


Figure 5: Direct Observation versus Gossip under a longer lifespan, $l = 10000$, $e = 0.7$, the presence of an error rate in gossip means that over time it can be outperformed by direct observation. Points plotted are means with 95% confidence interval calculated over ten replications.

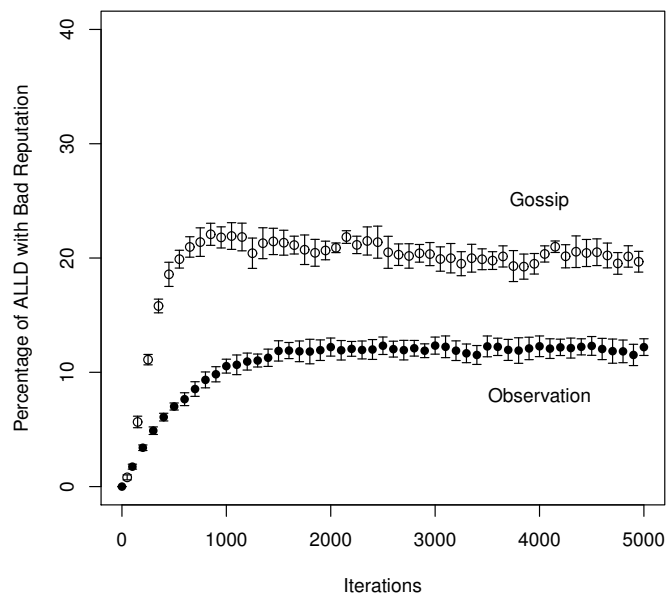


Figure 6: Direct observation versus gossip under a shorter lifespan $l = 3000$). Despite a ceiling determined by the error rate $e = 0.7$ here gossip can perform better than direct observation. This is because knowledge from observation will on average be acquired on a lower number of individuals per turn, therefore it suffers higher impact from the lower number of opportunities per lifetime. Thus the error generated during social transmission does not necessarily entail that direct observation is a better strategy. Points plotted are means with 95% confidence interval calculated over ten replications.

343 Results and Discussion

344 From Figure 5 it can be seen that, for the conditions of e and t in our experiments, gossiping individuals
345 can initially learn more quickly than those that rely on personal experience. This is in keeping with the
346 results demonstrated in Experiment 2. However transmission error acts as a ceiling on the maximum
347 amount of knowledge acquired from gossip—after initial progress a stage is reached where updates
348 are no longer improving on the information that an individual already holds. Learning by direct
349 observation may be slower, but since it is not subject to transmission error it can eventually deliver a
350 greater amount of information to each individual. However, this effect depends on lifespan which limits
351 what one agent can discover. Figure 6 shows populations competing under the same conditions except
352 for a shorter lifespan. Both strategies show a reduction in performance as a more changeable ecology
353 places greater demands on information-gathering abilities and reduces the relevance of information
354 held. However the two strategies do not degrade equally—for this set of parameters, the advantage of
355 accuracy over relevance has been overturned. Gossiping individuals have access to more information
356 as the error caused in transmission is outweighed by the speed advantage of gossip.

357 When lifespan is limited, relevant reputational information levels are reduced in four ways:

- 358 1. a record of reputation is useless when the individual it refers to no longer exists.
- 359 2. when individuals die the information they carry dies with them.
- 360 3. newborn individuals do not have a history of interactions on which their reputation can be based.
- 361 4. newborn individuals when entering into an interaction do not have the benefit of a history of
362 personal experience or gossip to draw on when choosing which move to play.

363 If information spreads more quickly by gossip than it can by direct observation, then gossip meets
364 these challenges more effectively, providing only that the rate of error is sufficiently low compared to
365 the relative rate of transmission. We formalise this provision below, and show results over a larger
366 range of parameters.

367 Formal Analysis of Gossip versus Direct Observation

368 The utility of gossip and experience can be formally analysed. By comparing the ability of Direct Ob-
369 servers and Gossipers to negatively repute ALLD agents, we can analyse the contexts where Gossiping
370 is a dominant strategy.

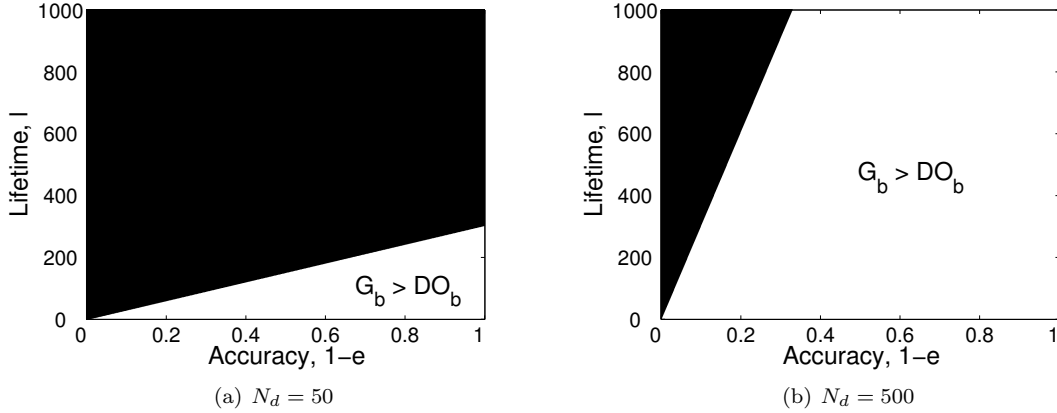


Figure 7: White represents the contexts where Gossipers diagnose more ALLD agents compared to Direct Observers. Analysis is run over a parameter sweep of lifespan (l) and error rate (e). Because lifespan does not scale with the size of the population, results vary with the latter. The number of ALLD agents (N_d , held constant at $1/3N$) in figures shown is **(a)** $N_d = 50$; ($N = 150$), and **(b)** $N_d = 500$; ($N = 1500$).

371 What is the probability that a Direct Observer will negatively repute an ALLD agent? As before,
 372 each Direct Observer begins with the belief that all agents are good. It is only through experience that
 373 agents attribute bad reputation. We define l as the Direct Observer's life expectancy. l denotes the
 374 number of agents it will meet during the course of its life. For simplicity and without loss of generality,
 375 we presume that at every interaction the agent meets a previously unmet agent, thus again maximising
 376 the benefit of the Direct Observation strategy. Therefore at the end of its life, the agent will have met
 377 l agents — one for each interaction.

378 During an interaction, the agent will meet an ALLD with probability P_d , where P_d is the percentage
 379 of ALLD agents in the population. Since the agents are judging others by direct experience, we assume
 380 no error in the observation. By the end of its life these optimal Direct Observers will have correctly
 381 classified lP_d ALLD agents as bad. The fraction of negatively reputed ALLD agents is this total
 382 divided by the number of ALLDs in the population. We define N_d as the number of ALLD agents in
 383 the population.

384 We presume that at any given time the age of the agents are homogeneously spread between the
 385 range $[1...l]$. Thus, the average age of the population is $\frac{l}{2}$. If DO_b is defined as the average fraction of
 386 negatively reputed ALLDs classified by Direct Observers, then it can be written as:

$$DO_b = \frac{lP_d}{2N_d} \quad (5)$$

387 In contrast to Direct Observers, Gossipers help each other classify ALLDs. As a consequence,
 388 while still constrained by l , the Gossipers benefit from a multiplier t which is the number of individ-
 389 uals they learn about on each cycle. However, they must contend with the possibility of erroneous
 390 communication.

391 Again for the benefit of simplicity, we assume a well-mixed population, such that in the first round,
 392 each ALLD is observed by at least one Gossiper. This also assumes that there are more Gossipers than
 393 ALLD, a safe assumption if half of the population is a Gossiper and a third ALLD. The Gossipers
 394 then relay the information to others. The likelihood that a Gossiper negatively reputes an ALLD
 395 is calculated as per Equation 4. The only undefined parameter from the equation is the number of
 396 rounds the agent has experienced, i . If each agent can experience l interactions in its lifetime, then, on
 397 average, a Gossiper will have experienced $\frac{l}{2}$ interactions. $i = \frac{l}{2}$ is then the average number of rounds
 398 the Gossiper has experienced. If G_b defines the chance that a Gossiper negatively reputes an ALLD,
 399 then:

$$G_b = (1 - e')(1 - (1 - t')^{\frac{l}{2}}) \quad (6)$$

400 (Note that $(1 - t')$ is raised to $\frac{l}{2}$, one half of lifespan.)

401 Now we can calculate when Gossipers will diagnose more ALLD agents than Direct Observers.
 402 Namely, when is:

$$G_b > DO_b \quad (7)$$

403 a somewhat complex relationship that depends on the interactions between e' , t' and l

$$e' < 1 - \frac{lP_d}{2N_d * (1 - (1 + t')^{l/2})}. \quad (8)$$

404 Results and Discussion

405 Figure 7 shows when Gossipers outperform Direct Observers for differing values of lifespan, $l =$
 406 $[1, 1000]$, and error rate, $e' = [0, 1]$. As in the computer model, 1/3 of the population is comprised
 407 of ALLD agents (i.e. $P_d = 1/3$). The transmission rate is held static at $t' = 0.5$. As explained in
 408 Experiment 2, because $i = l/2$, i is typically large enough that the transmission rate does not affect
 409 the fraction of negatively reputed ALLDs.

410 Because Direct Observation relies on personal experience over a fixed lifespan, the winning strategy

411 is not only dependant on lifespan and error rate, but also on the absolute number of ALLD agents, which
412 in this case scales with the size of the overall population. Figure 7(a) shows that direct observation
413 is often the best strategy when there are fifty ALLD agents in a population of one hundred fifty.
414 Even here, Gossipers can outperform Direct Observers if the accuracy of communication is sufficiently
415 high and if the lifespan of the agents is fairly short. In contrast, Figure 7(b) illustrates that the
416 Gossiping strategy is often preferred when there are five-hundred ALLD agents. Increasing the size of
417 the population decreases the chance of the Direct Observer strategy prevailing. In contrast, the Direct
418 Observer strategy does well when communication is highly erroneous or agents live a long time (cf.
419 Equation 8).

420 Direct Observer success depends on the number of ALLD agents because for total knowledge they
421 must individually experience each agent. As the size of the population grows, the Direct Observer
422 must spend much longer to categorize the same fraction of agents. In contrast, the Gossip strategy
423 can more easily scale to growing populations. Because the entire population works together to uncover
424 defectors, their success is much less perturbed by population growth. In summary, we have shown that
425 the success of gossiping compared to direct observation is dependent not only on lifespan, the rate of
426 communication and its fidelity, but also on the size of the population.

427 **3 General Discussion**

428 We have replicated in a spatial model the Nakamura and Masuda (2011) results and confirmed their
429 analysis, that gossip is able to sustain cooperation even when the information transmitted contains a
430 high degree of uncertainty. We propose an explanation for this: when assessing the value of information
431 in an evolutionary context we must incorporate not only accuracy but also relevance. We tested this
432 in an experiment which examined how transmission rate interacts with transmission error. We found
433 that socially-learned information even with a high incidence of error may outperform experience-based
434 learning if the individuals whose reputation is learned by the latter method are not the same as those
435 the agent subsequently encounters. In this we made the assumption that gossip is always as fast as or
436 faster than individual learning.

437 Finally we have demonstrated that the relative importance of accuracy and relevance is highly
438 sensitive to the ecology of the group, where by ‘ecology’ we mean demographic and life-history char-
439 acteristics determining network diffusion properties. We do not argue that gossiping individuals will
440 always outperform those that learn from direct experience. Our conclusion is that even if the incidence

441 of error in gossip is high, which is a reasonable assumption given that gossip is second-hand information
442 and talk is cheap, this does not mean that gossip is necessarily an inferior source of information.

443 A possible objection to this claim is that if changes in a society's ecology increase the frequency
444 of re-meeting between individuals then the scope to offset the error in gossip by its superior relevance
445 is reduced and learning by experience might prove more reliable (Roberts, 2008). Equation 8 shows
446 that in fact either technique may be preferred, depending also on the context of transmission rate,
447 transmission fidelity and lifespan. Gossip could still be a means of defence against rogue individuals
448 that travel further. Parochialism might be another defence, but this would also limit opportunities for
449 beneficial exchange between communities (Abbink et al., 2012; Rand et al., 2014).

450 The potential for increasing the probability of re-meetings will also vary according to the interaction
451 in question. Some types of interaction naturally take place very often during an individual's lifespan,
452 while others occur much less frequently. Hence as the results of Experiment 3 indicate, gossip may be
453 more relevant for interactions that are infrequent in the context of an agent's lifespan.

454 Importantly, we have gone to lengths to give Direct Observation perhaps unreasonable levels of
455 advantage in these simulations. In the case where perception of events was uncertain even for the
456 immediate observer (as often occurs in human socialisation), then if transmission fidelity is higher
457 than perception fidelity, it could be possible to *improve* the accuracy of information by combining
458 information based on the perception of multiple sources (Fahle and Poggio, 1981; Moravec, 1988;
459 Jacobs et al., 1991; Kao and Couzin, 2014). If for example learning by experience were to involve
460 direct observation as opposed to having been directly involved in the exchange, then we are able to
461 question the assumption that $k = 1$. This is true even for direct exchanges if the values of the costs
462 and benefits are for any reason obscured to those involved, a condition common in both treaties and
463 commerce.

464 Note that our results do not impact on the issues faced in explaining cooperation *per se*, for example
465 the objection that gossip itself assumes cooperation and that therefore using it to explain cooperation
466 is a regress. Of course, the invasion of individually maladaptive but socially beneficial traits has
467 been well explored and established elsewhere (e.g. Čače and Bryson, 2007; Alizon and Taylor, 2008).
468 However, our principle interest here is not so much to explain the ubiquitous presence of cooperation.
469 Here we take gossip and cooperation as a setting in which we have addressed the claim that cheap,
470 unreliable communication could not supplant more reliable direct observation.

471 4 Conclusion

472 Intuitively we tend to value accurate sources of information over inaccurate ones. Under the influence
473 of this intuition the evolution of socially-acquired information seems hard to explain as its accuracy
474 is likely to be lower than information acquired by direct experience. We have shown that the value of
475 an information source cannot be determined by its accuracy alone, relevance must also be taken into
476 account. Once this is done it is possible to explain that despite even a high level of inaccuracy, an
477 individual's best information-gathering strategy may still be to depend on gossip.

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