

Does discussion make crowds any wiser?

Hugo Mercier, Nicolas Claidière

▶ To cite this version:

Hugo Mercier, Nicolas Claidière. Does discussion make crowds any wiser?. Cognition, 2021, pp.104912. 10.1016/j.cognition.2021.104912. hal-03447665

HAL Id: hal-03447665 https://hal.science/hal-03447665v1

Submitted on 24 Nov 2021

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

1	DOES DISCUSSION MAKE CROWDS ANY WISER?
2	
3	Mercier, H.
4	Institut Jean Nicod,
5	Département d'études cognitives,
6	ENS, EHESS, PSL University, CNRS,
7	Paris France
8	hugo.mercier@gmail.com
9	ORCID ID : 0000-0002-0575-7913
10	
11	Claidière, N.
12	Aix Marseille University, CNRS, LPC, FED3C, Marseille, France.
13	nicolas.claidiere@normalesup.org
14	ORCID ID : 0000-0002-4472-6597
15	
16	Correspondance to Nicolas Claidière, Laboratoire de Psychologie Cognitive, 3 Place Victor Hugo, 13331
17	Marseille. Nicolas.claidiere@univ-amu.fr
18	
19	

20	Abstract
21	Does discussion in large groups help or hinder the wisdom of crowds? To give rise to the
22	wisdom of crowds, by which large groups can yield surprisingly accurate answers,
23	aggregation mechanisms such as averaging of opinions or majority voting rely on diversity of
24	opinions, and independence between the voters. Discussion tends to reduce diversity and
25	independence. On the other hand, discussion in small groups has been shown to improve the
26	accuracy of individual answers. To test the effects of discussion in large groups, we gave
27	groups of participants ($N = 1958$ participants in groups of size ranging from 22 to 212; mean
28	59) one of three types of problems (demonstrative, factual, ethical) to solve, first individually,
29	and then through discussion. For demonstrative (logical or mathematical) problems,
30	discussion improved individual answers, as well as the answer reached through aggregation.
31	For factual problems, discussion improved individual answers, and either improved or had no
32	effect on the answer reached through aggregation. Our results suggest that, for problems
33	which have a correct answer, discussion in large groups does not detract from the effects of
34	the wisdom of crowds, and tends on the contrary to improve on it.
35	
36	Keywords: Group decision making, wisdom of crowds, aggregation, majority rule, social
37	learning
38	Word count: 6259 (including references)
39	
40	

41

Does Discussion Make Crowds Any Wiser?

42

43 Ancient Athens is famous for its reliance on democratic decision making. Laws were put forward by a council of 500, and voted by an assembly of 6000 citizens. Judicial decisions 44 45 were made by courts of 200 jurors (Hansen, 1999). In each case, the assembled citizens would 46 listen to the arguments of the different parties, and the issue would be resolved by a simple 47 majority vote. Crucially, during these votes, discussion among citizens was not formally 48 allowed. Was this a wise rule? If answering this question might have helped Athenians make 49 better decisions, the generalization of democratic decision making means it is an even more 50 pressing question today. Crowds—defined here as any large group, whether or not they are 51 organized—play an increasingly important role, whether in politics—from mass protests to 52 citizens' assemblies—in the creation and diffusion of knowledge—from scientific consortia to 53 Wikipedia contributors—or in business, as companies try to make the best of their workforce's knowledge. 54

55 We start by reviewing arguments suggesting that discussion might hinder the wisdom 56 of crowds, and thus that groups might be better off aggregating their answers without 57 discussion, before turning to arguments suggesting instead that discussion might improve the 58 individual performance of the group members, without taking away the added value of the 59 wisdom of crowds. In the absence of empirical evidence directly bearing on this issue, we conduct a large-scale experiment in which 1958 participants in 33 groups with size ranging 60 from 22 to 212 participants (mean 59), are confronted with a variety of problems, first without 61 62 being able to discuss them, and then with discussion allowed. When an objective benchmark 63 for performance is available, our results suggest that discussion consistently improves 64 individual answers, and also often improves the answer reached through the wisdom of 65 crowds.

In ancient Athens, rules limiting discussion between citizens before a vote were no 66 doubt linked to the practical necessity of making a decision in a limited time frame (often half 67 a day) (Manville & Ober, 2003). More recently, theoretical work has suggested that these 68 constraints might have been wise, maximizing the chances that the citizens would vote for the 69 70 best available alternative. The most fundamental result underpinning the efficacy of majority 71 voting is the Condorcet Jury Theorem (Condorcet, 1785). For a dichotomous choice, the theorem "states that the probability that a majority votes for the better alternative exceeds p 72 73 [the probability that each voter selects the right option] and approaches 1 as *n* [the number of 74 voters] goes to infinity" (Ladha, 1992, p. 34). The efficacy of majority voting has been 75 demonstrated not only in models (e.g., Austen-Smith & Banks, 1996; Ladha, 1992), but also a 76 variety of experiments (e.g., Hastie & Kameda, 2005).

77 For the Condorcet Jury Theorem to apply, a set of constraints has to be respected— 78 that the voters are more likely than chance to vote for the best alternative, that they do not vote strategically, and, crucially here, that their decisions are independent of one another. If 79 80 some voters imitated others, without thinking for themselves, the effective size of the 81 assembly would be reduced, along with the chances that the majority supports the best 82 alternative. During discussion, voters are likely to influence each other, thereby potentially losing some of their independence, and lessening the benefits of majority voting (although 83 84 see, Estlund, 1994).

Besides majority voting, the other main phenomenon responsible for the wisdom of crowds is averaging. At least since Galton (1907), it has been well established that measures of central tendency such as the mean typically have a lower error than the mean individual error. For instance, when considering a range of numerical estimates that deviate more or less from a correct answer, the error of the mean answer will always be either lower than the mean error (if the correct answer is within the range of all the answers provided), or the same as the

mean error (otherwise) (see, e.g., Larrick & Soll, 2006). Moreover, for many distributions of
answers, the error of the mean is uncannily small compared to the mean error, a phenomenon
which has allowed averaging to improve performance on a variety of problems ranging from
political predictions to medical diagnoses (Surowiecki, 2005).

95 As in the case of majority voting, the risks of discussion for the benefits of averaging 96 are clear. During discussion, individuals are likely to converge on a middle of the road 97 answer, eliminating the most extreme views, which will reduce the diversity and the range of 98 answers, and lower the potential benefits of averaging. Even increases in individual accuracy 99 might not compensate for this loss of diversity (see, e.g., Hahn et al., 2019; Hong & Page, 100 2004; Lorenz et al., 2011). There are therefore good grounds to believe that discussion might 101 hamper information aggregation in large groups, which are most likely to benefit from the 102 wisdom of crowds. Indeed, the problem might be particularly acute in the type of densely 103 connected topologies that we will study here (Hahn et al., 2020).

104 By contrast, other results suggest that discussion might play a positive role. Small-105 group discussion has been shown to improve the average performance of the group members 106 on a wide range of problems, ranging from logical tasks to political predictions (e.g., Mellers 107 et al., 2014; Moshman & Geil, 1998; Trouche et al., 2014; for reviews, see, Laughlin, 2011; 108 Mercier, 2016; Mercier & Sperber, 2017). In some cases, discussion can even lead to answers 109 that are superior to those reached by any of the group members (e.g., Laughlin et al., 2003). 110 The question remains open of whether this improvement in performance, typically observed 111 in groups of at most five people (although see, Hastie et al., 1983, Hans, 2007 for 12-person juries, with less clearly correct answers, and Mellers et al., 2014 for larger groups interacting 112 113 through an internet forum), would translate to larger groups, which make discussion less 114 natural (Fay et al., 2000; Krems & Wilkes, 2019), and which might create more opportunities 115 for herding, or for the majority to impose its view regardless of its accuracy (e.g., Asch,

116 1956).

117 Still, it is possible that the improvement in performance yielded by small-group 118 discussion might also be observed in larger groups (on the difficulty for accurate answers to 119 spread widely, see, Moussaïd et al., 2017). Improvements in individual performance might 120 then be sufficient to compensate for the decrease in the diversity and independence of the 121 answers, such that discussion will improve, or at least not deteriorate, the wisdom of crowds 122 (be it obtained through majority voting, averaging, or other means of aggregation).

123 A few studies have tested whether discussion is detrimental to the wisdom of crowds 124 in large groups. In an experiment, mi-sized groups of participants (N = 12) had to make 125 numerical estimates (about, e.g., the population size in a city), and some participants were provided with the average group answer, and an opportunity to revise their estimate on that 126 127 basis (Lorenz et al., 2011). Although the average performance of these participants improved, 128 several indicators of the strength of the wisdom of crowds decreased (e.g. the degree of 129 diversity within the answers). Another study confirmed that receiving the average answer 130 from other participants leads to a decrease in diversity, but it also found that, for some 131 network configurations, the increase in individual accuracy more than compensated for this 132 loss of diversity (Becker et al., 2017). Importantly, in this latter experiment the participants 133 received the average group answer, but they were expressively forbidden from discussing 134 with one another. Several studies have shown that the increases in accuracy following 135 discussion are substantially larger than those following mere exposure to others' opinion (e.g., 136 Liberman et al., 2012; Minson et al., 2011). This experiment might thus underestimate the benefits of discussion. 137

138 In another experiment, a very large crowd (N = 5180) also had to provide numerical 139 estimates of various quantities (Navajas et al., 2018). Crowd members were then provided 140 with the opportunity to talk to each other in small groups (N = 5), for a very short amount of

time (1 min), and to revise their initial answers on the basis of this discussion. In this case, discussion had an unambiguously positive effect, as it increased not only individual performance, but also the answer reached through the wisdom of crowds. However, this study relied on the well-established improvement in performance following small-group discussion, and does not directly address the question of whether a broader discussion within the crowd would also yield such positive effects.

147 To the best of our knowledge, the study that most directly tested the effect of 148 discussion in medium sized groups (N = 11 to 25) used the following method—which we 149 describe in greater details, since it is similar to the method of the present experiments 150 (Claidière et al., 2017). In each group, participants were seated together in a room, following 151 a grid pattern. The participants were shown a logical or mathematical problem to solve, and 152 given five minutes to attempt to find an answer on their own. Participants then either had 153 fifteen minutes to talk about the problem with their neighbors (Discuss Condition), or to see 154 the response of their neighbors, without discussion (Silence Condition). Every minute, 155 participants recorded their answers, which allowed measuring changes in the percentage of 156 correct answers with time. After the initial five minutes of solitary reasoning, performance 157 improved faster in the Discuss than in the Silence condition. Moreover, a reanalysis of these 158 data shows that discussion vastly improved on the ability of the wisdom of crowds (here, 159 majority voting) to select the best answer. At the end of the first phase of solitary reasoning, 160 the correct answer was supported by the majority of the participants in only 3 out of 12 161 groups, while it was supported by the majority in all groups after discussion.

Even if this latter study shows that discussion improve individual answers and the aggregated answer yielded by the wisdom of crowds, it has several limitations. The group size, while larger than that used in most experiments on group decision making, was still modest. The problems used were known to yield massive improvement with small-group

166 discussion (Trouche et al., 2014). The participants were a homogenous group of students.

167 Finally, only one method of aggregating opinions—majority voting—was tested. A measure

168 of central tendency, for instance, might be more sensitive to a loss of diversity following

169 discussion (Hong & Page, 2004; Lorenz et al., 2011).

170 This overview of the literature suggests that there is no clear existing answer to the 171 question of whether large groups are better off discussing before their opinions are 172 aggregated. To start answering this question, we took advantage of a science festival, the 173 European Researchers' Night which would be attended by hundreds of people across 11 174 towns in France. In each town, a room was set up in which participants could take part in the 175 present experiment, as an introduction to research. As in the Discuss condition of Claidière et 176 al. (2017), after being presented with a problem, participants had five minutes to think about 177 it on their own, before being able to discuss it with their immediate neighbors for 15 minutes, 178 with their answers being recorded every minute.

We used three types of problems. First, two *demonstrative problems*, one of which being the bat and ball from the Cognitive Reflection Test (Frederick, 2005). Demonstrative problems have a solution that can be conclusively demonstrated using shared knowledge (Laughlin & Ellis, 1986). These problems constitute an extension and a replication to large, more diverse groups, of the experiment described above (Claidière et al., 2017).

Second, we used two *factual problems*, drawn from Navajas *et al.* (2018), such as "How many goals were scored in the XXX world cup?" If small-group discussion has been shown to improve performance on such problems (Navajas et al., 2018; Sniezek & Henry, 1989), the effects of large-group discussion, and the repercussions of the discussion for the value of the wisdom of crowds, have not been established to the best of our knowledge. Finally, we used two *ethical problems*, drawn from (Thorndike, 1937), such as "How much money should be awarded to compensate someone who lost a little finger in a

191	workplace accident?" Discussion in small groups on such problems typically does not lead to	
192	systematic changes of mind (Mercier et al., 2017). We did not expect that large-group	
193	discussion would lead to different outcomes. As a result, these problems were used as a	
194	control in which we did not expect discussion to have any systematic effect on the answers.	
195	If we expect the effects of small-group discussion to also be observed in large groups	
196	(as in Claidière et al., 2017 for demonstrative problems), we can derive the following	
197	hypotheses:	
198		
199	H1a For demonstrative problems, discussion improves performance more than solitary	
200	thinking.	
201	H1b For factual problems, discussion improves performance more than solitary	
202	thinking.	
203	H1c For ethical problems, discussion does not have a larger impact than solitary	
204	thinking.	
205		
206	When it comes to demonstrative problems, previous results also lead to the prediction	
207	that discussion will improve both individual performance and the aggregate answer.	
208		
209	H2 For demonstrative problems, discussion leads to better aggregate answers, as	
210	selected through majority voting.	
211		
212	By contrast, for factual problems, it is unknown whether the loss of diversity and	
213	independence will compensate for any potential individual gain in accuracy. As a result, we	
214	formulate the following research question:	
215		

216	RQ1 For factual problems, does discussion lead to worse, equivalent, or better
217	aggregate answers, as selected through averaging?
218	
219	Method
220	Participants
221	The experiment was part of the European Researchers' Night, a pan-European science
222	fair organized by researchers to introduce the public to the world of science and research. In
223	France, the organizing committee of the 2017 edition gave us the opportunity to organize a
224	large participative experiment involving 11 cities and 1958 participants (1048 females).
225	Participants were visitors to the science fair, who came in a large room to take part in an
226	experiment advertised as not being suitable for children younger than 12 (90% of participants
227	reported an age between 13 and 60; median = 24). There were two to six consecutive groups
228	in each city (totaling 33 groups ranging from 20 to 208 individuals [mean 58]), which led to a
229	total of between four to seven groups (259 to 468 participants) per problem. More details can
230	be found in the ESM.
231	
232	Materials
233	The six problems we used as material were:
234	
235	Paul and Linda (demonstrative problem 1). Paul looks at Linda; Linda looks at John; Paul is
236	married; John isn't married; Is someone married looking at someone who isn't married?
237	Answers provided: Yes [correct] / No / We can't tell.
238	
239	Bat and Ball (demonstrative problem 2). A candy and a baguette cost 1.10€ together. The
240	baguette costs 1€ more than the candy. How much does the candy cost? <i>Correct answer:</i>

241	0.05€.
242	
243	World Cup (factual problem 1). How many goals were scored in the football world cup of
244	2010? Correct answer: 145
245	
246	Elevators (factual problem 2). How many elevators are there in New York's Empire State
247	Building? Correct answer: 73
248	
249	Little Finger (ethical problem 1). How much money should be awarded to compensate
250	someone who lost a little finger in a workplace accident?
251	
252	Worms (ethical problem 2). How much money should be awarded to compensate someone
253	who finds they have been eating earthworms in their restaurant meal?
254	
255	Procedure
256	The experiment took part in large rooms with chairs arranged in a grid pattern. As
257	participants arrived, they were asked to sit close to each other so that their seating
258	arrangement would be as close as possible to a square grid, with no empty seats. Once
259	everyone was seated, a trained researcher explained to the participants that they were taking
260	part in a real experiment, that they could leave the room at any time, that their anonymous
261	data would be used in a scientific publication and that by giving us their response sheet at the
262	end of the experiment they agreed to these conditions.
263	Answer sheets were distributed that contained 15 rows, one row for each time step,
264	with the space for an answer to the problem, some demographic questions that were answered
265	immediately (group number, seat number, town, age, gender), and a white space for free

writing. After a brief explanation of the Silence Phase of the experiment, and the importance
of not talking, showing each other their answers, or using their phones to check the answer,
the experiment started. The problem was displayed on a large screen so that all participants
could start answering it at the same time. After 20s, the participants provided their first
answer. Four more answers were gathered at succeeding 1-min intervals.

271 Participants were then told that they would now be able to discuss their answers with 272 their neighbors (Discussion Phase). Neighbors were defined as the eight (maximum) 273 participants surrounding them. Participants were told that the goal was for them to reach a 274 consensus. After they were given the signal to start discussing, the participants had to write 275 down their answer every minute, as in the Silence Phase, for 10 minutes. Time was kept by 276 the experimenter who prompted everyone to write down their answer every minute. At the 277 end of the experiment a 15 min debrief explained the state of the art in group decision 278 making, the purpose of the experiment, and the hypotheses. Participants were also encouraged 279 to advertise the experiment to other potential participants at the fair, but without revealing its 280 purpose and proceedings. Importantly, we changed problems between the groups in each city 281 in order to make sure that participants were completely naïve (i.e. even if they had been 282 informed by a previous participant, they would face a different problem).

283

284 Data coding and analysis

Response sheet for demonstrative and factual problems were coded using a crowdsourcing platform. Three independent coders coded the responses of each participant and when available the modal response was retained. In cases in which three different coders disagreed, often due to mistyping from the coders, the experimenters returned to the original response sheet to determine the most likely response (less than 1% of the responses were reevaluated).

Regarding ethical problems, that required more judgment, one independent coder coded all responses from the 499 participants using four categories (for Little finger: a number, a monthly allowance, cost of medical intervention and other; for Worms: a number, the price of the meal, medical costs, and other).

295

296 Data exclusion and response variable

297 We excluded a total of 11% of responses from analysis. This percentage varied 298 between problems, but, crucially, it did not vary with time (see ESM for detailed table). For 299 Paul and Linda, we excluded responses that were not any of the three proposed options (<1%) 300 and used as response variable a binary variable with 1 for correct response and 0 for any of 301 the other two responses. For Bat and Ball, we excluded responses that were not 5 or 10 cents 302 (6%) and used a similar binary variable, with 1 for correct response and 0 for the incorrect 303 response. For the Elevators and World Cup problems we excluded responses that were not 304 numeric, and responses above the 99% quantile to avoid extremely large values (such as 305 "123456"; 7% and 12% of data were excluded resp.).

306 Finally, for the Worms and Little Finger problems, we excluded data from the "other" 307 category (25% and 27% resp.) and re-coded responses as a binary response variable with 1 308 being the most frequent response at the end of the Silence Phase (i.e. the majority option 309 before discussion) and 0 for all alternative responses. We should note, however, that our 310 ethical problems, which had no correct answer, were quite different from the other problems 311 and raised a number of issues, such that no strong conclusion can be drawn from them. Based 312 on the advice of reviewers we decided to present the results of the ethics problems in the ESM 313 only.

314

315 Statistical method

Analysis were carried out using R (R Core Team, 2020), mixed models were analyzed
with the package lme4 (Bates et al., 2015) and ggplot2 was used to produce the figures
(Wickham, 2016).

319

320 Data availability

321 All the data analyzed here are available at DOI: 10.17605/OSF.IO/CFWV2

- 322
- 323

Results

324 To test H1a, b, and c, we sought to determine whether discussion had a larger effect on 325 the answers than solitary reflection. Figure 1 summarizes the evolution through time of the 326 different groups with the average response for each problem (Supplementary videos 1 to 6 327 illustrate this evolution using the spatial layout of the rooms in which the experiment was 328 carried out; the videos of each group are available in the public repository of the experiment). 329 Following Claidière et al. (2017), we used mixed models to study the interaction between the 330 experimental phase (Silence vs. Discussion), and time during the first 10 timesteps (to 331 maintain the same number of observations in the two phases: 5 in each of the Silence and 332 Discussion phases). We report the models that combined the problems of each type; however, 333 we also analyzed each problem independently and found that the results of the combined 334 models also applied to each problem independently (full reporting of the models can be found 335 in the Electronic Supplementary Materials). As in our previous study we found that discussion 336 favored the dissemination of the correct response for the two demonstrative problems ($\beta =$ 337 0.38, SE = 0.04, z = 8.37, p < 0.001). For the two factual problems, there was also a 338 significant interaction between the Silence and Discussion phases, with a reduction in the 339 distance to the correct response observed only during the Discussion Phase ($\beta = -2.31$, SE = 340 0.74, df = 6586, t = -3.12, p = 0.002; see Fig.2).

341

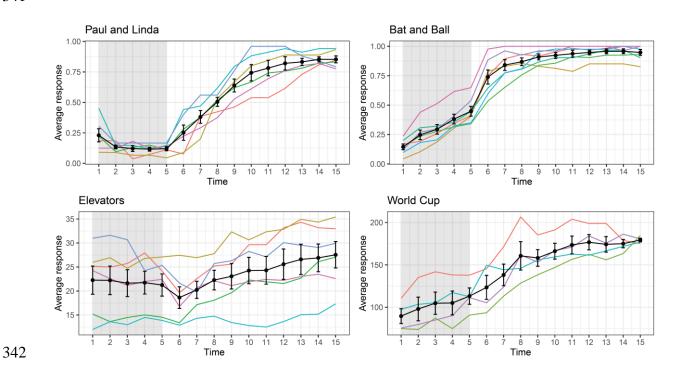
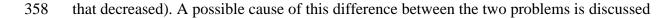


Figure 1: Evolution of the group response for each problem through the Silence (shaded area) and Discussion phases. Each colored line represents a unique group mean response and the black line represents the between group mean (+/- SE). The correct answer to the Elevators problem was 73 and to World Cup problem 145.

To test H2, and answer RQ1, we turn to the effect of discussion on the aggregate answers. For demonstrative problems, we find that discussion leads to better aggregate answers. At the end of the Silence Phase, out of 13 groups, only two had a majority of correct responses (both for the Bat and Ball). By contrast, all groups had a majority of correct responses at the end of the Discussion Phase (a significant improvement, McNemar's chisquared = 9.10, df = 1, p = 0.003).

For factual problems we found that the error of the mean response (how the mean response in each group differed from the correct answer) decreased for the Elevators problem (Fig. 2; all six groups had a lower error of mean at the end of the Discussion Phase compared to the end of the Silence Phase; binomial test, p = 0.03). By contrast, there was no evidence of a decrease for the World Cup problem (two groups had a value that increased and two a value



below.

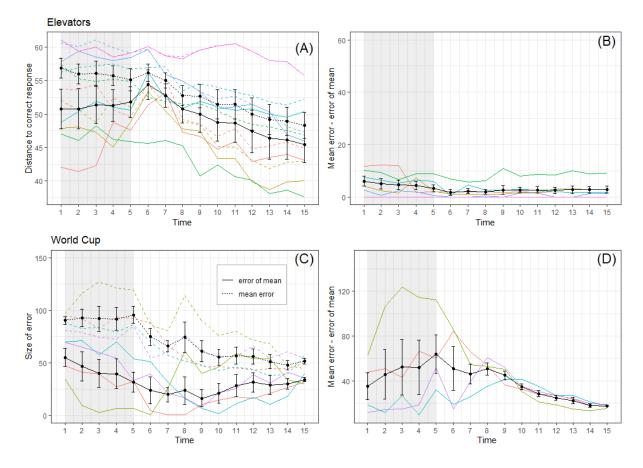


Figure 2: Effect of discussion on the wisdom of crowds. Evolution of the mean error made by individuals (the mean of all the individual errors) and the error of the mean response (mean responses which are depicted in Figure 1) (A, C), as well as the difference between the two (B, D) for the Elevators and World Cup problems. Each colored line represents a unique group mean response and the black line represents the between group mean (+/- SE).

366

360

Discussion

367 H1a, b, and c were confirmed. For both demonstrative and factual problems,

368 discussion improved performance over solitary thinking. The results also clearly supported

369 H2: for demonstrative problems, discussion improved not only individual answers, but also

- the answers favored by majority voting, which went from two correct answers at the end of
- beginning of the Discussion Phase, to 13 out of 13 at the end.

372 Regarding RQ1, the answer is more equivocal. For one factual problem (Elevators),

discussion consistently improved not only on individual answers, but also on the answers
reached through averaging within each group. By contrast, for the other factual problem
(World Cup), discussion improved on individual answers, but not on the answers reached
through averaging.

377 To understand the differential impact of discussion on the wisdom of crowds in the 378 two factual problems, it is useful to go back to Figure 2. As noted previously, the mean error 379 of the participants decreased through time for both problems. Moreover, the wisdom of 380 crowds effect was present throughout the experiment, with the error of the mean being always 381 inferior to the mean error of individuals (Fig. 2A, 2C). However, while the size of the gain 382 through aggregation (i.e. the difference between the mean error and the error of the mean) 383 stayed relatively constant during the Discussion Phase for Elevators (Fig. 2B), it decreased for 384 World Cup (Fig. 2D).

385 To make sense of this difference, we can consider two ways for the mean error to 386 decrease: (i) if most answers are distant from the correct answer, and there is a directional 387 shift towards the correct answer, or (ii) if most answers aren't too distant from the correct 388 answer, and there is a reduction of the variance in the answers, with the most extreme answers 389 converging towards the correct answer. Overall, in Elevators, there is no decrease in variance 390 (Fig. 2A), but there is a general shift towards the correct answer, which the overwhelming 391 majority of participants had initially underestimated (Fig. 1). By contrast, in World Cup, there 392 is no directional shift towards the correct answer, with the average answer being as distant 393 from the correct answer at the beginning than at the end of the discussion (Fig 1); however, 394 there is a reduction in the variance of the answers (Fig. 2C). Such a reduction in variance 395 lowers the mean error, but not the error of the mean, thereby decreasing the difference 396 between the two.

397

It is also worth noting that in all but one of the 10 groups facing factual problems, on

398 average participants moved more towards the correct answer than towards what was the 399 average group answer at the beginning of the discussion (see ESM, Table S3, and Fig. S2). 400 Indeed, on the whole participants barely moved towards the average answer (Elevators, 1.34; 401 World Cup, 0.10), but they consistently moved towards the correct answer (Elevators, 7.30; 402 World Cup, 36.03). This means that the improvement observed during discussion did not 403 result from participants simply converging towards an answer corresponding to the average at 404 the beginning of the Discussion Phase, as might be expected if participants felt the pull of the 405 majority (see, e.g., Moussaïd et al., 2013). Instead, in every group participants moved towards 406 the correct answer. For factual problems (as for logical problems), in the course of discussion 407 participants appear to have been pulled by arguments towards the correct answer (see, 408 Claidière et al., 2017; Mercier & Sperber, 2017). 409 Conclusion 410 Are crowds wiser with or without discussion? The literature makes conflicting 411 predictions, and to answer this question we gave groups of medium to large size (N = 20 to 412 208) a problem to tackle individually first, and then through discussion with their neighbors.

413 When there were objective benchmarks, individual answers consistently improved with

414 discussion, while aggregate answers improved in most cases and never consistently worsened.

415 When it comes to problems for which a correct answer exists, our results strongly 416 argue in favor of discussion. First, for the four problems with correct answers studied here— 417 two logical, demonstrative problems, and two factual problems-discussion always improved 418 the mean individual answer. Second, in three out of four cases, discussion led to better 419 aggregate answers, aggregated either through majority voting (the two demonstrative 420 problems), or through averaging (one factual problem). Third, in the last case with no 421 improvement in aggregate answers, discussion was not detrimental to the aggregated answer 422 because it had no effect. Thus, discussion had no detrimental effect on the wisdom of crowds

423 for the problems examined here.

424 Our results also demonstrate the effectiveness of discussion in a more qualitative 425 manner. For the two demonstrative problems, 15 minutes of discussion yielded enormous improvements in individual answers, which moved from 12% correct to 84% correct for Paul 426 427 and Linda, and from 41% correct to 91% correct for the Bat and Ball. Remarkably, in the case 428 of Paul and Linda, all groups reached at least 75% of correct answers, even though they had 429 started with at best 17%. These results thus demonstrate the robustness of the 'truth-wins' 430 scheme, by which a single individual with a correct answer to a demonstrative problem can 431 convince a group, since we also observe its effects in large and diverse groups.

432 The positive effects of discussion are also clear for the two factual problems. In the 433 Elevator problem, all groups correctly increased their average answer through discussion, 434 moving from a mean error of 55 at the beginning of the discussion to a mean error of 48 at the 435 end. In the World Cup problem, discussion nearly halved the mean error from 96 to 52. We 436 also note that asking participants to estimate the number of goals scored in one specific world 437 cup is a very high bar and it is remarkable that the average number of goals scored in the past 438 six world cups is 160 goals, a difference of only 19 goals with the grand average reached at 439 the end of the discussion. Moreover, in our experiments, participants were constrained in 440 terms of who they could discuss the problems with. Giving people flexibility in network 441 formation might further increase the advantages of discussion (see, e.g., Almaatoug et al., 442 2020). Alternatively, constraining networks to optimize the flow of information has also been 443 shown to improve accuracy when discussion is not possible, but the same results might extend 444 to situation in which discussion is possible (Jönsson et al., 2015).

Our results have theoretical and practical consequences. They support theoretical
frameworks that postulate the power of discussion to change minds for the best (Mercier &
Sperber, 2011, 2017), and they show that the loss in independence and diversity in the

answers during discussion can be largely compensated by the increase in accuracy, contrary to
what had been suggested (e.g., Hong & Page, 2004; Lorenz et al., 2011). Practically, our
results show that discussion is a robust tool to improve not only individual, but also collective
answers, even in large and diverse groups, at least for problems that have a correct answer.

453 Acknowledgments

We thank all the staff of the French European Researchers' Night, the local and the national 454 455 organizing committees and in particular Matteo Merzagora and Lionel Maillot for their help 456 in setting up the "Grande Experience Participative". We also thank all the people that 457 conducted the experiment, Sacha Altay, François Druelle, Justine Epinat, Annabelle Goujon, 458 Julie Gullstrand, Sébastien Lerique, Heather McLeod, Mathilde Menoret, Virginie Postal-Le 459 Dorse, Jean-Pierre Thibaut, Romain Trincherini, and Jean-Baptiste Van der Henst. The 460 authors declare no competing interests. NC gratefully acknowledges financial support from ASCE (ANR-13-PDOC-0004) and LICORNES (ANR-12-CULT-0002). HM gratefully 461 462 acknowledges financial support from FrontCog (ANR-17-EURE-0017), and PSL (ANR-10-463 IDEX-0001-02).

464

4	<u> </u>
4	65

References

466 Almaatouq, A., Noriega-Campero, A., Alotaibi, A., Krafft, P. M., Moussaid, M., & Pentland,

467 A. (2020). Adaptive social networks promote the wisdom of crowds. Proceedings of the

- 468 *National Academy of Sciences*, *117*(21), 11379–11386.
- 469 Asch, S. E. (1956). Studies of independence and conformity: A minority of one against a

470 unanimous majority. *Psychological Monographs*, 70(9), 1–70.

- 471 Austen-Smith, D., & Banks, J. S. (1996). Information aggregation, rationality, and the
- 472 Condorcet jury theorem. *American Political Science Review*, 90(01), 34–45.
- 473 Bates, D., Mächler, M., Bolker, B., & Walker, S. (2015). Fitting Linear Mixed-Effects Models
- 474 Using lme4. Journal of Statistical Software, 67(1), 1–48. https://doi.org/10.18637/jss.v067.i01
- 475 Becker, J., Brackbill, D., & Centola, D. (2017). Network dynamics of social influence in the
- wisdom of crowds. *Proceedings of the National Academy of Sciences*, *114*(26), E5070–
 E5076.
- 478 Claidière, N., Trouche, E., & Mercier, H. (2017). Argumentation and the diffusion of counter-
- 479 intuitive beliefs. Journal of Experimental Psychology: General, 146(7), 1052–1066.
- 480 Condorcet. (1785). Essai sur l'application de l'analyse à la probabilité des décisions rendues
- 481 à la pluralité des voix. L'imprimerie royale.
- 482 Estlund, D. (1994). Opinion leaders, independence, and Condorcet's jury theorem. *Theory*
- 483 *and Decision*, *36*(2), 131–162.
- Fay, N., Garrod, S., & Carletta, J. (2000). Group discussion as interactive dialogue or as serial
 monologue: The influence of group size. *Psychological Science*, *11*(6), 481–486.
- 486 Frederick, S. (2005). Cognitive reflection and decision making. *Journal of Economic*
- 487 Perspectives, 19(4), 25–42.
- 488 Galton, F. (1907). Vox populi. *Nature*, 75(7), 450–451.
- 489 Hahn, U., Hansen, J. U., & Olsson, E. J. (2020). Truth tracking performance of social
- 490 networks: How connectivity and clustering can make groups less competent. *Synthese*,
 491 *197*(4), 1511–1541.
- Hahn, U., von Sydow, M., & Merdes, C. (2019). How communication can make voters
 choose less well. *Topics in Cognitive Science*, *11*(1), 194–206.
- 494 Hans, V. P. (2007). Deliberation and dissent: 12 angry men versus the empirical reality of
- 495 juries. Chi.-Kent L. Rev., 82, 579.
- 496 Hansen, M. H. (1999). The Athenian democracy in the age of Demosthenes: Structure,
- 497 *principles, and ideology.* University of Oklahoma Press.
- 498 Hastie, R., & Kameda, T. (2005). The robust beauty of majority rules in group decisions.
- 499 *Psychological Review*, *112*(2), 494–50814.
- 500 Hastie, R., Penrod, S., & Pennington, N. (1983). Inside the Jury. Harvard University Press.
- 501 Hong, L., & Page, S. E. (2004). Groups of diverse problem solvers can outperform groups of
- 502 high-ability problem solvers. Proceedings of the National Academy of Sciences of the United
- 503 States of America, 101(46), 16385.
- Jönsson, M. L., Hahn, U., & Olsson, E. J. (2015). The kind of group you want to belong to:
- 505 Effects of group structure on group accuracy. *Cognition*, *142*, 191–204.
- 506 Krems, J. A., & Wilkes, J. (2019). Why are conversations limited to about four people? A
- theoretical exploration of the conversation size constraint. *Evolution and Human Behavior*,
 40(2), 140–147.
- 509 Ladha, K. K. (1992). The Condorcet jury theorem, free speech, and correlated votes.
- 510 American Journal of Political Science, 617–634.
- 511 Larrick, R. P., & Soll, J. B. (2006). Intuitions about combining opinions: Misappreciation of
- 512 the averaging principle. *Management Science*, *52*, 111–127.
- 513 Laughlin, P. R. (2011). *Group problem solving*. Princeton University Press.

- 514 Laughlin, P. R., & Ellis, A. L. (1986). Demonstrability and social combination processes on
- 515 mathematical intellective tasks. *Journal of Experimental Social Psychology*, 22, 177–189.
- 516 Laughlin, P. R., Zander, M. L., Knievel, E. M., & Tan, T. S. (2003). Groups perform better
- than the best individuals on letters-to-numbers problems: Informative equations and effective reasoning *Lournal of Parsonality and Social Psychology* 85, 684, 694
- 518 reasoning. Journal of Personality and Social Psychology, 85, 684–694.
- Liberman, V., Minson, J. A., Bryan, C. J., & Ross, L. (2012). Naïve realism and capturing the "wisdom of dyads." *Journal of Experimental Social Psychology*, 48(2), 507–512.
- 521 Lorenz, J., Rauhut, H., Schweitzer, F., & Helbing, D. (2011). How social influence can
- 522 undermine the wisdom of crowd effect. *Proceedings of the National Academy of Sciences*,
- *523 108*(22), 9020–9025.
- 524 Manville, B., & Ober, J. (2003). A Company of Citizens: What the World's First Democracy
- 525 Teaches Leaders About Creating Great Organizations. Harvard Business Review Press.
- 526 Mellers, B., Ungar, L., Baron, J., Ramos, J., Gurcay, B., Fincher, K., Scott, S. E., Moore, D.,
- 527 Atanasov, P., Swift, S. A., & others. (2014). Psychological strategies for winning a
- 528 geopolitical forecasting tournament. *Psychological Science*, 25(5), 1106–1115.
- 529 Mercier, H. (2016). The argumentative theory: Predictions and empirical evidence. *Trends in*
- 530 *Cognitive Sciences*, 20(9), 689–700.
- 531 Mercier, H., Castelain, T., Hamid, N., & Marín Picado, B. (2017). The power of moral
- 532 arguments. In J. F. Bonnefon & B. Trémolière (Eds.), *Moral Inferences*. Psychology Press.
- 533 Mercier, H., & Sperber, D. (2011). Why do humans reason? Arguments for an argumentative
- theory. Behavioral and Brain Sciences, 34(2), 57–74.
- 535 Mercier, H., & Sperber, D. (2017). *The Enigma of Reason*. Harvard University Press.
- 536 Minson, J. A., Liberman, V., & Ross, L. (2011). Two to Tango. *Personality and Social*
- 537 *Psychology Bulletin*, *37*(10), 1325–1338.
- 538 Moshman, D., & Geil, M. (1998). Collaborative reasoning: Evidence for collective rationality.
- 539 *Thinking and Reasoning*, *4*(3), 231–248.
- 540 Moussaïd, M., Herzog, S. M., Kämmer, J. E., & Hertwig, R. (2017). Reach and speed of
- judgment propagation in the laboratory. *Proceedings of the National Academy of Sciences*,
 114(16), 4117–4122.
- 543 Moussaïd, M., Kämmer, J. E., Analytis, P. P., & Neth, H. (2013). Social influence and the
- 544 collective dynamics of opinion formation. *PloS One*, 8(11), e78433.
- 545 Navajas, J., Niella, T., Garbulsky, G., Bahrami, B., & Sigman, M. (2018). Aggregated
- knowledge from a small number of debates outperforms the wisdom of large crowds. *Nature Human Behaviour*, 2(2), 126.
- 548 R Core Team. (2020). R: A language and environment for statistical computing. R
- 549 Foundation for Statistical Computing. https://www.R-project.org/
- 550 Sniezek, J. A., & Henry, R. A. (1989). Accuracy and confidence in group judgment.
- 551 *Organizational Behavior and Human Decision Processes*, 43(1), 1–28.
- 552 Surowiecki, J. (2005). The Wisdom Of Crowds. Anchor Books.
- 553 Thorndike, E. L. (1937). Valuations of certain pains, deprivations, and frustrations. *The*
- 554 *Pedagogical Seminary and Journal of Genetic Psychology*, 51(2), 227–239.
- 555 Trouche, E., Sander, E., & Mercier, H. (2014). Arguments, more than confidence, explain the
- 556 good performance of reasoning groups. Journal of Experimental Psychology: General,
- 557 *143*(5), 1958–1971.
- 558 Wickham, H. (2016). ggplot2: Elegant graphics for data analysis. Springer-Verlag.
- 559