Native Amazonian children forego egalitarianism in merit-based tasks when they learn to count.

DEVELOPMENTAL SCIENCE

UNCORRECTED PROOF

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Research Highlights

• In the Tsimane’, children who can count are more likely to produce merit-based distributions compared to children who cannot count.

• This difference is better predicted by children’s ability to count than by their age or years in school.

• These findings suggest that learning the logic of natural numbers and counting can influence social cognition.
Abstract

Cooperation often results in a final material resource that must be shared, but deciding how to distribute that resource is not straightforward. A distribution could count as fair if all members receive an equal reward (*egalitarian distributions*), or if each member’s reward is proportional to their merit (*merit-based distributions*). Here, we propose that the acquisition of numerical concepts influences how we reason about fairness. We explore this possibility in the Tsimane’, a farming-foraging group who live in the Bolivian rainforest. The Tsimane’ learn to count in the same way children from industrialized countries do, but at a delayed and more variable timeline, allowing us to de-confound number knowledge from age and years in school. We find that Tsimane’ children who can count produce merit-based distributions, while children who cannot count produce both merit-based and egalitarian distributions. Our findings establish that the ability to count—a non-universal, language-dependent, cultural invention—can influence social cognition.
Introduction

Fair distribution of resources is important in many aspects of human life, including cooperative tasks. However, deciding what counts as fair is not straightforward. Very young children prefer egalitarian distributions, in which all members get an equal share of the resources (e.g., Schmidt & Sommerville, 2011; Sommerville, Schmidt, Yun, & Burns, 2012). In doing so, they frequently ignore merit (Damon, 1975), need (Huntsman, 1984), and group membership (Fehr, Bernhard, & Rockenbach, 2008; Olson & Spelke, 2008). As they grow older, children consistently start producing and preferring more complex distribution methods based on effort and other factors (Almås, Cappelen, Sørensen, & Tungodden, 2010; Huntsman, 1984; Nelson & Dweck, 1977; Damon, 1975, 1980; Sigelman & Waitzman, 1991; Lisi, Watkins, & Vinchur, 1994). This developmental change may be driven by children’s increased experience in cooperative contexts, by explicit pedagogy, or by a mixture of both.

Here we propose an additional dimension that may influence how children reason about resource distributions: the acquisition of the logic of natural numbers (i.e., the logic that sets have an exact size that can be counted, and the words and mechanisms for calculating these sizes via counting). Knowing how to count and understanding its logic can affect how people distribute resources in several ways. Most obviously, if a child wants to give someone an exact number of objects, he or she needs to be able to count up to that number first. However, children not only learn how to count to calculate a set’s

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1 Although the literature does not converge on an exact developmental milestone when children forego egalitarianism, the studies consistently report that younger children are more likely to behave as egalitarians in the presence of merit, need, and social affiliation differences.
exact size; they also learn the numerical concepts underlying counting. For example, young children do not realize that substituting one element in a set (removing one element and putting in another) leaves the set’s exact size constant, nor that adding or taking one element away does not (Izard, Streri, & Spelke, 2014; see also Davidson, Eng, & Barner, 2012).

In light of what children learn, mastering the logic of natural numbers may affect both how we think about fairness and how we act upon these beliefs. To illustrate this, consider a simple scenario: A child has to distribute a set of cookies between two people, one of whom worked harder than the other. Despite the simplicity of the scenario, taking merit into account is not trivial. First, the child needs to translate a work or effort difference into a merit difference (i.e., the person that worked harder deserves a bigger reward). Next, the child needs to transform the abstract merit difference into a concrete reward difference (i.e., based on the merit difference, how many more cookies does the harder-working person deserve?). Last, the child needs to be able to distribute the cookies based on these earlier judgments. Pre-numerical children may succeed in the first step and believe that the harder working person deserves a bigger reward, but they may nevertheless have trouble with the last two steps. Specifically, if children are concerned with producing justifiable distributions, then not understanding how simple manipulations affect a set’s size may impede them from deciding how to transform a merit difference into a reward difference. At an even broader level, mastering exact number concepts may be a precursor for children’s understanding that almost anything can be exactly or approximately quantified (including merit). This is in fact crucial for
utilitarianism (e.g., Bentham, 1879; Mill, 1906), one of the most influential theories on fairness, which relies on the assumption that anything can be quantified even when how to quantify it is unclear. In the face of these impediments, children may default to a fairness rule that they can comprehend and consistently produce: egalitarianism (See also Damon, 1975 for a similar proposal on the relation between numerical knowledge and social cognition.)

Exploring the effect of numerical competency on fairness judgments is difficult in industrialized cultures because both the ability to count and the ability to understand the logic of counting correlate with age (Izard, Streri, & Spelke, 2014; Lipton & Spelke, 2006; Piantadosi, Jara-Ettinger, & Gibson 2014). However, this is not true in the Tsimane’. The Tsimane’ are a non-industrialized farming-foraging group living in the Bolivian Amazon (Huanca, 2008). The Tsimane’ are an ideal population for testing how learning the logic of the natural numbers may influence fairness judgments. They learn to count in the same manner as children in industrialized countries do, but on a delayed and more variable timeline (Piantadosi, et al., 2014). Thus, this population allows us to disentangle the role of natural number understanding from other maturational skills that correlate tightly with age in industrialized cultures. To discover how mastery of the logic of natural numbers and counting affects merit-based fairness judgments, we ran a simple resource distribution task with the Tsimane’.

**Experiment**

**Methods**
Participants. We recruited 70 children (mean age: 6.53 years; SD: 1.93 years; range 3-12; 38 males, 32 females) from six Tsimane’ communities near San Borja, Bolivia. All work was done in collaboration with the Centro Boliviano de Investigación y de Desarrollo Socio Integral (CBIDSI), which provided interpreters, logistical coordination, and expertise in Tsimane’ culture.

Procedure. Children’s ages and years of education were gathered through parental reports and, when available, school records. Children’s ability to count was assessed through the Give-N task (e.g., Wynn, 1990). In this task, one sheet of paper with 10 chips was placed next to a second sheet of paper. Children were first asked to move 4 out of the 10 chips from one sheet to the other. We next followed a staircased procedure in which a higher number was requested whenever the child moved the correct number of chips and a lower number was requested whenever the child moved an incorrect number of chips. This procedure continued until either (1) the child successfully moved eight chips from one sheet to the other (thus having shown that he or she could produce the appropriate number of objects when asked for four, five, six, seven, and eight objects), (2) the child’s ability to count could be determined after the first eight queries using the classification rules from Piantadosi, et al., (2014), or (3) the child wanted to stop. Occasionally, participants’ performance made it impossible for us to assess the knower-level stage through the pre-determined staircasing procedure (for example, if a participant always moved three chips, the staircasing procedure would oscillate between requesting four chips and three chips). When this happened the experimenter restarted the
staircasing procedure at a smaller query, thus allowing us to test if the participant could count up to lower numbers.

The first and last author, blind to the participant’s demographic information (age and years of education) and to their performance in the fairness task (described below), independently coded each participant as *pre-numerical* or as a *full counter* based on their performance on the Give-N task. That is, each coder independently determined if each participant’s errors were more likely due to performance and distraction errors (in which case they were coded as *full counters*) or if their errors were consistent enough to believe that the child did not understand counting (in which case they were coded as *pre-numerical*). Overall, the two coders agreed on 98.5% of the trials (Cohen’s Kappa inter-rater agreement = 0.968; $p < 0.0001$). Because mastery of counting and mastery of number concepts are tightly related in the Tsimane’ (Jara-Ettinger, Piantadosi, Spelke, & Gibson, *under review*), the pre-numerical versus full counter distinction roughly separated participants into children who cannot count nor understand number concepts and children who can count and have mastered numerical concepts.

Next, children completed the fairness task. Participants saw two cutout drawings of two identical children, differentiable only by their shirt color, placed on opposite sides of the table. The interpreter explained that one day, the two children had been sent to pick bananas. The first child worked very hard and brought back many bananas. The second

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2 In the number literature, pre-numerical children are often called subset-knowers and full counters are called CP-knowers. Subset-knowers can be further divided into a sequence of ordered stages (0- to 4-knowers; Piantadosi, et al., 2014). However, our interest here is on what happens when children master numerical concepts, so there was no need to determine the exact stage of the pre-numerical children.

3 Cases in which the first two coders disagreed were resolved by the second author following the methodology of Solomon, Johnson, Zaitchik, & Carey (1996).
child did not work very hard and only brought back a few bananas. As the interpreter described their performance, the experimenter placed a picture of 18 bananas (arranged in a 6 by 3 matrix) next to the hard-working child and a picture of four bananas (arranged in a 2 by 2 matrix) next to the non-hard-working child. The position of the two children relative to the participant and the order in which they were introduced were randomized across children.

To ensure that the scenario was clear, the interpreter asked participants to point to the child who had worked the hardest. The interpreter then told the participants that the children would receive some cookies as a prize for collecting the bananas. Two conditions varied the number of cookies that each participant distributed: Children were given either four cutout pictures of cookies (small-set condition, \( N = 35 \)) or ten cutout pictures of cookies (large-set condition, \( N = 35 \)). The interpreter asked each participant to distribute the cookies across the two children. All participants distributed their entire set of cookies across the two children, and the experimenter recorded the number of cookies that each participant gave to the harder-working child. Participants completed only a single trial to keep the experiment simple because the Tsimane’ participants were unaccustomed to experiments.

**Results.** Participants who failed to identify the harder-working child in the inclusion question were excluded from further analysis (\( N = 9 \) of 70 participants, 12.86%). We categorized the remaining participants as *merit users* if they gave more cookies to the harder-working child, or as *egalitarians* if they gave an equal amount of cookies to each
child. Because our goal was to compare egalitarians to merit-users, we excluded five participants who gave fewer cookies to the harder-working child and thus fit into neither category ($N=5, 7.14\%$). The remaining dataset contained 56 participants; 29 in the four-cookie condition and 27 in the ten-cookie condition. Table 1 shows the full data from the experiment.

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4 Four of these five participants were pre-numerical children in the large-set condition and one participant was a full counter in the small-set condition. See Table 1 for demographic information. Including these participants as merit users, egalitarians, or members of their own separate category does not affect our results.
Table 1: Results from the experiment. Each row shows a participant’s performance. The table shows each participant’s age (Age column), years in school (School column), the set-size they were given to distribute (Set-size column), whether they could count or not (Full counter column), and the number of cookies they gave to the harder-working child (Answer column). Children who took merit into account (i.e., when the answer is higher than half the set’s size) are color-coded in light gray. The bottom part of the table (dark gray) shows participants who gave fewer cookies to the harder-working child and were thus unclassifiable.

To test our main prediction—that children’s ability to count affects whether their distributions reflect an appreciation of merit—we divided participants into children who could count (full counters; \( N = 23 \)) and children who could not (pre-numerical; \( N = 33 \)) (see Procedure). Figure 1 shows the pattern of responses. As predicted, the proportion of merit-using participants was higher among full counters. 57.58% of pre-numerical children produced egalitarian distributions (19 out of 33 participants) and 42.42% produced merit-based distributions (14 out of 33 participants). In contrast 26.09% of full counters produced egalitarian distributions (6 out of 23 participants) and 73.91% produced merit-based distributions (17 out of 23 participants). Furthermore, a non-parametric permutation test revealed that the proportion of merit-based distributions was significantly higher among full counters, compared to pre-numerical children (\( p < 0.05 \)).
Figure 1: Children who cannot count (pre-numerical) show no overall bias between egalitarian and merit-based distributions. In contrast, children who can count (full counters) are biased towards producing merit-based distributions, giving more cookies to the harder working child. Each point shows the proportion of children making each choice and the vertical bars show 95% confidence intervals on the estimate.

Next, we tested if the difference between children who produced egalitarian and merit-based distributions could be explained by differences in their ages or years in school. Figure 2 shows each participant’s distribution type as a function of their age and education. If merit-usage were determined purely by age, or another factor tightly correlating with age, then merit-based distributions should be clustered on the right of the figure and egalitarian distributions should be clustered on the left of the figure. If merit-usage were determined by years in school, or any experience tightly correlating with years in school, then merit-based distributions should be clustered on the top of the figure and egalitarian distributions should be clustered on the bottom of the figure. Last, if the
distribution type could be explained by some interaction between age and years in school, then merit-based distributions should be clustered in one (or more) of the corners of the figure. However, Figure 2 reveals none of these patterns. Instead, merit-based distributions are interspersed among egalitarian distributions, suggesting that the difference between egalitarian and merit-using children is not due to age, years in school, or a combination of these two. Consistent with this, results from a multiple logistic regression found no significant effect of age, years in school, or their interaction ($p>0.69$ in all cases).\(^5\)

![Figure 2: Participants’ choice of distribution as a function of their age (x-axis) and their years in school (y-axis). Red circles represent participants who produced an egalitarian distribution and blue circles represent participants who produced a merit-based distribution. Merit-based distributions are not biased towards the right (implying age matters), to the top (implying education matters), or clustered on the top right (implying...](image)

\(^5\)To ensure this null result was not because age and years of education correlate, we next performed two separate logistic regressions predicting distribution type from years of education and age. Neither revealed a significant effect: $\beta=0.27 \ (p=0.20)$ per year of education and $\beta=0.09 \ (p=0.54)$ per year in age.
a combination of age and education matters). Instead, they are intermixed with egalitarian distributions, suggesting that, in the Tsimane’, foregoing egalitarianism in merit-based tasks is not the result of age or exposure to school.

Next, to disentangle how age, counting ability, and set size affect children’s use of merit, we performed a logistic multiple regression with merit-vs-egalitarian as the binary dependent variable. Age (z-scored), children’s understanding of number (pre-numerical or full counter), set size (as a sum-coded factor), and the interaction of the latter two, were input as predictors (independent variables). Table 2 shows the results from the regression. Consistent with our main finding, only children’s ability to count yielded a significant influence on fairness, controlling for the other factors, suggesting that children’s use of merit is strongly guided by their ability to count and not by their age.

|               | Estimate | Std. Error | z-value | Pr(>|z|) |
|---------------|----------|------------|---------|----------|
| Intercept     | -0.3873  | 0.3860     | -1.003  | 0.316    |
| Age           | -0.1579  | 0.3826     | -0.413  | 0.680    |
| Set-size      | 0.4281   | 0.3679     | 1.164   | 0.245    |
| Full counter  | 1.5350   | 0.7200     | 2.132   | 0.033*   |
| Set-size:Full counter | -0.4615 | 0.6231     | -0.741  | 0.459    |

Table 2: Results from a multiple logistic regression with children’s use of merit as the dependent variable. Children’s age (z-scored), their ability to count, set-size condition (as a sum-coded factor), and the interaction of set-size with counting ability were input as predictors.
The qualitative pattern of our results also appeared in each of the two set-size conditions when analyzed separately. In the small-set condition (distributing four cookies), the percentage of merit-based distributions increased from 52.94% among pre-numerical children (9 out of 17 participants) to 75% among full counters (9 out of 12 participants). Similarly, in the large-set condition (distributing ten cookies), the percentage of merit-based distributions increased from 31.25% among pre-numerical children (5 out of 16 participants) to 72.73% among full counters (8 out of 11 participants). However, a quantitative analysis looking at each condition separately revealed a significant effect in the large-set condition ($p=0.048$ by permutation test) but not in the small-set condition ($p=0.2764$ by permutation test). Together with the overall analyses, these findings suggest that children’s acquisition of number may have a stronger influence when the set is large.

**Discussion**

Here we have shown that in the Tsimane’, a farming-foraging group living in the Amazonian region of Bolivia, a child’s ability to count is a strong predictor of whether their distribution of resources reflects an appreciation of merit. Full counters were significantly more likely to produce a merit-based distribution compared to pre-numerical children. Our results show that mastery of number—a non-universal, cultural invention (Frank, Everett, Fedorenko, & Gibson, 2008)—can influence merit-based fairness judgments.
Critically, our findings could not be explained by age or years of education. Thus, it is unlikely that our findings could be due to children foregoing egalitarianism as a function of exposure to social activities (which increases with age and occurs during school). Nevertheless, the ability to produce merit-based distributions may not be related to the acquisition of number concepts, but simply guided by some underlying factor that correlates with counting ability, but not age or school attendance.

In principle, children could have relied on approximate magnitudes and distributed the cookies by estimating the size of different piles. Thus, if they wanted, participants should have been able to give more cookies to the harder working child without counting, especially in the small set condition. 42.42% (N=14) of pre-numerical children gave more cookies to the harder-working child, suggesting they relied on approximate magnitudes to produce this distribution. Thus, the fact that full counters were significantly more likely to use merit, regardless of their age, may suggest that children prefer exact, justifiable methods in scenarios that involve fairness. That is, children may prefer reproducible rules in which they can exactly determine how many more cookies the harder-working agent received. Further work is needed to explore these possibilities.

Our results do not imply that children begin producing merit-based distributions immediately after they master counting. In our task, we could only assess whether a child was a full counter or not, but we do not know for how long full counters had mastered

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6 Note that pre-numerical children should have trouble even if the reward difference is a small number (e.g., “One more cookie for the harder-working child.”) because they do not understand how simple manipulations, such as adding one element, change a set’s size (Izard et al., 2014).
number concepts. In addition, our results also do not imply that the change towards merit-based distributions is only caused by the acquisition of counting. Children may also undergo a conceptual change in their ideas of fairness, which becomes more powerful when they become able to manipulate sets with high precision.

Why do pre-numerical children produce egalitarian distributions more often than full counters do? Is it because they do not understand merit (not believing that the child who worked harder deserves more)? Three recent studies suggest this is not the case. Baumard, Mascaro, and Chevallier (2012) found that if a task makes it impossible for participants to produce an egalitarian distribution (for example, by having a small and a big cookie, or an odd number of cookies), three-year-old children (who may understand counting, but are likely too young to understand the logic of natural numbers; Davidson, Eng, & Barner, 2012) give more resources to whomever contributed the most towards the task. Similarly, Kanngiesser and Warneken (2012) found that three- and five-year-olds can take merit into account even in first-party contexts. Along the same lines, Sloane, Baillargeon, and Premack (2012) showed that, before their second birthday, toddlers expect rewards to be distributed only among people who participated in the task for which they are being rewarded (i.e., they behave as believing that those who did nothing deserve nothing). Thus, although pre-numerical children tend to produce egalitarian distributions, they nevertheless seem to understand merit.

Given the evidence that pre-numerical children understand merit, why do only full counters consistently produce merit-based distributions? Is it because it’s easier for them
to decide how to transform an abstract merit difference into a concrete reward difference? Is it because they have less trouble producing the distribution they believe is the most fair? Or is it a combination of both? Intuitively, learning to count should influence children’s ability to produce the distributions they believe are fair. But more fundamentally, learning the underlying logic of numbers and counting—specifically, that sets have exact sizes—may influence how children transform merit differences into exact reward difference. Here, we used a single measure, the Give-N task, which roughly separates children into those who cannot count nor understand the logic of natural numbers, and those who can and do. Because these two acquisitions are tightly correlated in the Tsimane’ (Jara-Ettinger et al., under review), we do not know how learning numerical concepts and learning to count may independently influence children’s performance in merit-based fairness tasks. Further work would be needed to disentangle their roles.

Interestingly, the acquisition of counting and number concepts may extend beyond merit-based fairness tasks, as many tasks where fairness intuitions come into play involve transforming an abstract measure, such as need or group affiliation, into a concrete reward difference, which the child must then produce. Learning that sets can be quantified may lead children to focus on the idea that differences in merit, need, or need can also be quantified and mapped onto material reward differences.

In conclusion, by testing a population in which mastery of number concepts and counting are weakly related with age and years in school, we were able to isolate an effect of
number knowledge in how children distribute resources in merit-based fairness tasks. Our results show that numerical concepts can influence how we reason about fairness and they highlight the need for a theoretical framework integrating how numerical concepts empower our intuitive theories.

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