Sources of knowledge in children's acquisition of the successor function

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Abstract

Through understanding the successor function, i.e. for any natural number n, its successor is n+1, we can gain the insight that the natural numbers are infinite. Recent work has suggested acquisition of this logical property is more protracted than previously thought, with a fully generalized understanding of the successor function not apparent until 5.5 to 6 years of age. While such work links successor knowledge with counting mastery, the exact processes underlying this developmental transition remain unknown. Here, we examined two hypothesized mechanisms: (1) productive counting knowledge, or mastery of the recursive process through which number words are generated, and (2) formally trained arithmetic, specifically the '+1' operation. We tested the relationship between successor knowledge, productive counting, and arithmetic proficiency in 140 3.5 to 6 year-olds. We found that while both productive counting and arithmetic mastery predicted successor knowledge, mean arithmetic performance was significantly lower for all children, even those at ceiling in implementing the successor function This surprising dissociation suggests children do not draw upon the '+1' operation in acquiring the successor function. Rather, these findings are more consistent with the hypothesis that this knowledge is acquired through productive counting, and the recognition that numbers are recursively generated through an implementation of the successor function.

Keywords: Add your choice of indexing terms or keywords; kindly use a semi-colon; between each term.

Introduction

One of our most profound achievements as learners is the ability to extract limitlessly productive rules from limited data. While this capacity makes us prodigious learners in general, one of its most powerful applications is in the domain of number. Although linguistic number input and expression are undoubtedly finite, we nevertheless come to know the natural numbers to be infinite; for every number, its successor can be obtained by simply adding '1.' Further, we understand that this successor function can be endlessly implemented due to the recursive nature of symbolic number. How does such an understanding arise? In this work, we explored two hypothesized causal mechanisms (productive counting knowledge and formally trained arithmetic) by which children might acquire the successor function, a critical component in the logical framework of natural number.

Early in numerical development, children do not treat number as a productive form of representation. Although they may be able to recite the count list up to about 10 (Fuson 1988), they do not acquire the meanings of these numerals until much later, with one, two, and three effortfully acquired over the course of about 12 months (Wynn, 1992). Around the time that children acquire the meaning of four, however, they suddenly seem to understand the relationship between

their memorized count list and cardinality, referred to as the Cardinal Principle (CP).

Even after children acquire the CP, however, they may not yet understand that numbers are generated through a recursive process. Recent work has found evidence that children do not understand the successor function as a truly generalized property of natural number until several years after acquisition of the CP. While many children who have acquired the CP (CP-knowers) can implement the successor function for small numbers, such as 4 or 5 (Sarnecka & Carey, 2008), their ability to do so for any number is mediated by their counting proficiency (Davidson, Eng, & Barner, 2012; Cheung, Rubenson, & Barner, 2017). In this work, only children who demonstrate mastery of the count list are able to apply the successor function to any number queried. These results indicate that counting proficiency, rather than CP-knowledge, is implicated in acquisition of this logical property.

This observed correlation with counting mastery suggests two potential paths by which children might acquire the successor function. The first, which we refer to as productive counting knowledge, is that as children come to understand the recursive nature of number, they notice that the next number generated via that process is always done so by the addition of '1,' or an implementation of the successor function (Cheung et al., 2017). This hypothesis would therefore predict that children who specifically demonstrate knowledge of recursion are more likely to have acquired the successor function.

The second hypothesized mechanism is that children learn the successor function through explicit instruction, rather than an induction made over the count list. As children learn formal arithmetic operations, such as 4+1=5 and 5+1=6, it is possible that they hypothesize the '+1' rule to hold true for any number. On this account, we would expect to find that children's successor function knowledge is significantly related to mastery of these rote-learned 'math facts.' In the present work, we tested these two potential paths to successor function acquisition in 140 3.5- to 6-year-old CP-knowers (3.58 - 5.98, Mage = 4.9, SDage = 0.61,). We used a modified version of the Unit Task (Sarnecka & Carey, 2008) to test children's successor knowledge. In this task, an experimenter placed some number of fish on a board, and said, "Look! There are N fish in the pond." The experimenter then covered these fish with a lilypad, and placed one additional fish next to the lilypad, saying "Look! Are there N+1 or N+2 fish now?" Children received 16 trials, with N ranging from 6 to 95.

Method

Participants

We recruited 140 children between the ages of 3;6 and 5;11 (M = 4.89, SD = 0.61) out of a planned sample of 150.

Stimuli and Design

Procedure

First level headings should be in 12 point, initial caps, bold and centered. Leave one line space above the heading and 1/4~line space below the heading.

Results

Second level headings should be 11 point, initial caps, bold, and flush left. Leave one line space above the heading and $1/4^{\sim}$ line space below the heading.

Discussion

Third-level headings should be 10 point, initial caps, bold, and flush left. Leave one line space above the heading, but no space after the heading.

Formalities, Footnotes, and Floats

Use standard APA citation format. Citations within the text should include the author's last name and year. If the authors' names are included in the sentence, place only the year in parentheses, as in (1972), but otherwise place the entire reference in parentheses with the authors and year separated by a comma (Newell & Simon, 1972). List multiple references alphabetically and separate them by semicolons (Chalnick & Billman, 1988; Newell & Simon, 1972). Use the et. al. construction only after listing all the authors to a publication in an earlier reference and for citations with four or more authors.

For more information on citations in RMarkdown, see here.

Footnotes

Indicate footnotes with a number¹ in the text. Place the footnotes in 9 point type at the bottom of the page on which they appear. Precede the footnote with a horizontal rule.² You can also use markdown formatting to include footnotes using this syntax.³

Figures

All artwork must be very dark for purposes of reproduction and should not be hand drawn. Number figures sequentially, placing the figure number and caption, in 10 point, after the figure with one line space above the caption and one line space below it. If necessary, leave extra white space at the bottom of the page to avoid splitting the figure and figure caption. You may float figures to the top or bottom of a column, or set wide figures across both columns.

Two-column images

You can read local images using png package for example and plot it like a regular plot using grid.raster from the grid package. With this method you have full control of the size of your image. Note: Image must be in .png file format for the readPNG function to work.

You might want to display a wide figure across both columns. To do this, you change the fig.env chunk option to figure*. To align the image in the center of the page, set fig.align option to center. To format the width of your caption text, you set the num.cols.cap option to 2.

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Figure 2: One column image.

R Plots

You can use R chunks directly to plot graphs. And you can use latex floats in the fig.pos chunk option to have more control over the location of your plot on the page. For more information on latex placement specifiers see **here**

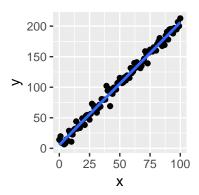


Figure 3: R plot

Tables

Number tables consecutively; place the table number and title (in 10 point) above the table with one line space above the caption and one line space below it, as in Table 1. You may float tables to the top or bottom of a column, set wide tables across both columns.

¹Sample of the first footnote.

²Sample of the second footnote.

³Sample of a markdown footnote.



Figure 1: This image spans both columns. And the caption text is limited to 0.8 of the width of the document.

You can use the xtable function in the xtable package.

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-0.03	0.09	-0.3	0.78
X	2.08	0.09	23.4	0.00

Table 1: This table prints across one column.

Acknowledgements

Place acknowledgments (including funding information) in a section at the end of the paper.

References

Chalnick, A., & Billman, D. (1988). Unsupervised learning of correlational structure. In *Proceedings of the tenth annual conference of the cognitive science society* (pp. 510–516). Hillsdale, NJ: Lawrence Erlbaum Associates.

Newell, A., & Simon, H. A. (1972). *Human problem solving*. Englewood Cliffs, NJ: Prentice-Hall.