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The log-linear response function of the bounded number-line task is unrelated to the psychological representation of quantity

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Keywords: Numerical Cognition; Number line; logarithmic function; log-linear shift; computational modeling

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Abstract

The bounded number-line task has been used extensively to assess the numerical competence of both children and adults. One consistent finding is that young children display a logarithmic response function, whereas older children and adults display a more linear response function. Traditionally, these log-linear functions have been interpreted as providing a transparent window onto the nature of the participants' psychological representation of quantity (termed here a *direct response strategy*). Here we show that the direct response strategy produces the log-linear response function regardless of whether the psychological representation of quantity is compressive or expansive. Simply put, the log-linear response function results from task constraints rather than the psychological representation of quantities. We also demonstrate that a proportion/subtraction response strategy produces response patterns that almost perfectly correlate with the psychological representation of quantity. We therefore urge researchers not to interpret the log-linear response pattern in terms of numerical representation.

The log-linear response function of the bounded number-line task is unrelated to the psychological representation of quantity

According to the National Council of Teachers of Mathematics (NCTM) Principles and Standards document ("*Common Core Standards*", 2010), the link between numerical symbols and one's psychological representation of quantity (termed *number sense*) is a crucial component in K-12 students' ability to understand symbolic numbers, learn new systems of number representations, understand relationships among numbers and number systems, and make reasonable estimates. The bounded number-line task is one of the principal tools for *teaching and measuring* number sense in children (see e.g., "*Common Core Standards*", 2010; Berteletti, Lucangeli, Piazza, Dehaene, & Zorzi, 2010; Dehaene, Izard, Spelke, & Pica, 2008; Laski & Siegler, 2007; Thompson & Opfer, 2008). Much of the evidence supporting the use of the bounded number-line task to measure number sense comes from papers making conclusions about the underlying psychological quantity representations from the logarithmic and linear response patterns of children and adults (e.g., Geary, Hoard, Nugent, & Byrd-Craven, 2008; Laski & Siegler, 2007; Opfer & Siegler, 2007; Siegler & Booth, 2004; Siegler & Opfer, 2003). In contrast to this well-established view, here we demonstrate that the logarithmic and linear response patterns in the bounded number-line task provide no information about the form of the underlying psychological representation of quantity.

The bounded number line task is very simple. A participant is presented with a number together with a horizontal line. The line designates a range of values, with the minimum value situated at the left-most edge of the line and the maximum value situated at the right-most edge. The participant is asked to indicate where on the line the corresponding quantity, conveyed by the number, would be located given the bounded range of values. Of central importance is the shape of the function (henceforth the *response function*) relating participants' quantity estimates to actual

target values. Various cognitive strategies have been put forward to explain performance. Traditionally, researchers have assumed that a participant's response function to the bounded number-line task provides a direct window on their psychological understanding of quantities (Booth & Siegler, 2006; Opfer & Siegler, 2012; Opfer, Thompson, & Kim, 2016; Siegler & Booth, 2004; we term this a *direct response strategy*). Researchers who accept the direct response strategy have interpreted the bounded number-line data as indicating that children's psychological understanding of quantities is logarithmic, whereas older children and adults psychological understanding of quantities is linear (Booth & Siegler, 2006; Laski & Siegler, 2007; Opfer & Siegler, 2005, 2012; Siegler & Opfer, 2003, Siegler & Booth, 2004; Siegler & Ramani, 2006; Siegler, Thompson & Opfer, 2009).

More recently, researchers have extended these ideas by fitting a log-linear function to the data. By introducing systematic changes to the parameters of this function, it can mimic a log function, a linear function, and a mixture of the two (Anobile, Cicchini, & Burr, 2012; Opfer, Thompson, & Kim, 2016). Despite this apparently promising approach, here we show that constraints of the bounded number-line task force a log-linear response pattern when the participant uses a direct response strategy regardless of the underlying numerical representation of quantity. We conclude, therefore, that researchers cannot draw any conclusions about participants' representation of quantity based on a log-linear (and thus a logarithmic) response pattern.

There is general agreement in the literature that the psychological understanding of a quantity, say a "5", depends on a representation captured by a distribution on a continuum of quantities (often described as a "mental number-line"). Different assumptions that have been made about the nature of these quantity distributions and their spacing on the line. Two important theoretical alternatives are (i) the linear model with scalar variability, and, (ii) the logarithmic model with fixed variability. The linear model with scalar variability assumes that the means of the distributions that represent one's psychological understanding of successive integers

(e.g., 1, 2, 3, etc.) are equidistant from one another, but the standard deviations (SDs) of those distributions increases proportionally with the quantity they represent. The logarithmic model with fixed variability assumes that the means of the distributions that represent the psychological understanding of successive integers are logarithmically spaced, but the SDs of those distributions remain constant.

The fact that young children produce a logarithmic response pattern in the bounded number-line task sits comfortably with the logarithmic model (Opfer & Siegler, 2005, 2012; Thompson & Opfer, 2008). In addition, the fact that older children and adults produce a near linear response pattern on the bounded number-line task sits comfortably with the linear model (Booth & Siegler, 2006; Siegler & Booth, 2004). This age-related pattern has led to the hypothesis that there is a development change such that an initial logarithmic representational system is replaced by a linear representational system as a person ages. Such a view is captured by *the log-linear shift hypothesis* (Siegler & Ramani, 2006; Siegler, et al., 2009).

The log-linear shift hypothesis is based on the assumption that the bounded number-line task provides a transparent window on the nature of the mental representation of quantities. It is assumed that a participant's response function will mirror his or her underlying representation of quantity. According to Siegler and Booth (2004), "Several groups of investigators have hypothesized that children's estimation reflects their internal representation of numbers" (p. 429). Furthermore, Opfer et. al (2016) made the argument that the bounded number-line provides an ideal template for directly measuring the underlying quantity representations, in the following way, "Because line length itself is not psychophysically compressive or expansive (Lu & Doshier, 2013), the task provides a relatively straightforward method for assessing compression in numerical magnitude representations." (p. 12).

Recently, an alternate view has emerged that focuses on the role that strategies play in the bounded number-line task (see e.g., Barth & Paladino, 2011; Barth, Slusser, Cohen, & Paladino, 2011; Chesney & Matthews, 2013; Cohen &

Blanc-Goldhammer, 2011; Cohen & Sarnecka, 2014; Friso-van den Bos, Kroesbergen, Van Luit, Xenidou-Dervou, Jonkman, Van der Schoot, et al., 2015; Hurst & Cordes, 2015; Hurst, Relander, & Cordes, 2016; Slusser, Santiago, & Barth, 2013). The counter argument is that (i) that successful completion of the bounded number-line task depends on the deployment of an appropriate cognitive strategy, and, (ii) only when these strategies are modelled, can a meaningful estimate of the participant's underlying quantity representation be obtained. For example, Cohen and Blanc-Goldhammer (2011) presented evidence that adults used a proportion/subtraction strategy to estimate the position of the target value on the bounded number-line. The strategy was put forward by Hollands and Dyre (2000) in their work on how observers make proportion estimates. According to Hollands and Dyre (2000), observers estimate the position of the target value from the left-hand boundary as well as from the right-hand boundary (e.g., upper bound – target value). This provides a means by which the observer scales the line to the quantities. This strategy is modelled by the Cyclic Power Model (CPM, see Hollands & Dyre, 2000). Cohen and Blanc-Goldhammer (2011) found that the CPM provided a superior fit to adult's bounded number-line data than either the linear or logarithmic functions. These results were replicated in older children (Barth & Paladino, 2011; Cohen & Sarnecka, 2014; Slusser et al., 2013).

To date, researchers have assumed that the direct response strategy and the proportion/subtraction strategy give rise to response functions that correlate with the underlying quantity representations. Perhaps surprisingly, this assumption has never been tested. As a consequence, we report on a number of computer simulations that test this critical assumption. The key issue is whether reasonably accurate estimates of underlying quantity representations can be made from the data produced from the direct response hypothesis and, separately, the proportion/subtraction hypothesis.

Methods

We ran computer simulations of the direct response hypothesis and the proportion/subtraction hypothesis.

Quantity representation

For all simulations, we first simulated an underlying quantity representation for the integers contained within the upper and lower boundaries of the number-line. Here, we only consider a number-line ranging from 0-50, but the results generalize to any range. For each integer, Θ_i we generated a quantity distribution such that the psychological representation of that integer was set to $\psi_i = \Theta_i + e_i$, where β denotes the numerical bias and e denotes the error around the mean. For the simulation, we varied β from 0.6 (negatively accelerating, similar to the log) to 1.4 (positively accelerating, similar to exponential), in 0.2 intervals. When $\beta = 1$, there is no bias. We simulated a variety of error distributions: that is, normal distributions with fixed and scalar variance, and the gamma distribution. We focus on the simulations with a gamma distribution (shape parameter = $\log(\text{target})+1$). The gamma distribution produces similar data to the normal distribution with scalar variance, but has the advantage of never producing a value below zero. Any variation in the outcomes was mostly attributable to the chosen standard deviation.

Task completion

Direct response strategy

For each target value, we sampled an estimate from the corresponding quantity distribution. Because the direct response strategy does not specify a means by which an observer maps the length of the line to quantity, the response estimate equaled the sampled estimate multiplied by a unit length. We varied unit lengths from 0.5-3.0 in 0.5 intervals. A unit length of 1.0 indicates an accurate mapping of line length to quantity. The response estimate was valid if fell between the lower and upper limits of the bounded number-line and invalid if it fell outside these limits. We

repeated the sampling and calculation process until we obtained a valid estimate. If we failed to calculate a valid response estimate in 1000 attempts, we assigned the response estimate the value of the boundary it was closest to. We repeated this process 20 times for each target value.

Proportion/subtraction strategy

For each target value, we sampled an estimate from its corresponding quantity distribution and one from the quantity distribution corresponding to the distance from the right boundary (upper boundary – target). We then calculated a response estimate from either the one-cycle or two-cycle CPM from those values. We repeated the sampling and calculation process until we obtained a valid estimate. If we failed to obtain a valid estimate in 1000 attempts, we assigned the response estimate the value of the boundary it was closest to. We repeated this process 20 times for each target value.

Results

For each strategy (direct response, one-cycle proportion/subtraction, and two-cycle proportion/subtraction) we fit the log-linear function developed by Anobile et. al, (2012; see also Cicchini, Anobile, & Burr, 2014). The log-linear function has two free parameters: (i) $p(\log)$ which identifies the proportion of log fit (vs. linear fit) and, (ii) a a scaling parameter. For the one-cycle proportion/subtraction, and two-cycle proportion/subtraction, we also fit the one- and two-cycle CPM. The CPM functions have one free parameter: β which is the estimate of numerical bias. We present the fit statistics in Tables 1 and 2. We present figures of the fits for a subset of the simulations assuming gamma distributions in Figures 1 and 2.

As seen in Table 1 and Figure 1, the log-linear function fit all direct response simulations well, regardless of the bias of the underlying quantity representation. In fact, when the bias of the underlying quantity representation is negatively accelerating (similar to a log function) the log-linear function fit least well ($\beta = 0.6$ condition). More concerning is the fact that the log-linear function fit extremely well (mean $r^2 = 0.89$) for all conditions in which the underlying quantity representation is

positively accelerating (expansive, rather than compressive). In addition, $p(\log)$ increases with the positively accelerating biases. These findings result from the fact that observers cannot respond outside the bounds of the bounded number line. As such, the bounds of the number-line truncate the distributions of the larger numbers, thus forcing a negatively accelerating response function. We therefore conclude that the direct response strategy in the bounded number-line task manifests as a log-linear response pattern regardless of the underlying psychological representation of quantity. The clear implication is that the log-linear pattern observed in behavioural data arise because of the constraints of the bounded number-line task, rather than the participant's psychological understanding of numbers.

As seen in Table 2 and Figure 2, the one and two-cycle functions fit all proportion/subtraction response simulations well (mean $r^2 = 0.986$), regardless of the bias of the underlying quantity representation. Importantly, the estimated β from both the one and two-cycle CPM functions are almost perfectly correlated with the simulated β of the quantity representations ($r > 0.99$), though with a small, but consistent, underestimation. Finally, Figure 2 shows the typical "M" shaped error pattern that results in the bounded number task emerges when participants use the half way mark as a reference point. One can conclude from these data that the proportion/subtraction response strategy in the bounded number-line task manifests as a CPM response pattern that reflects the response bias regardless of the underlying psychological representation of quantity.

Finally, we also note that the log-linear function fit all proportion/subtraction response simulations well (mean $r^2 > 0.98$), despite the fact that they were generated using the proportion/subtraction response strategy. This is because the proportion/subtraction response strategy produces relatively linear response patterns with slight curves that the log-linear function can mimic. This simulation underscores Hollands and Dyre (2000) caution that one must take extreme care not to confuse the linear (or log-linear) function for the CPM.

Discussion

The results of the simulations are very clear: log-linear (and thus logarithmic) response patterns in the bounded number-line task provide no information about the underlying quantity representation of the observer. Log-linear (and thus logarithmic) response functions result from the constraints of the task itself: the bounds of the number-line truncate the distributions of the larger numbers. In contrast, if participants are able to carry-out the proportion/subtraction strategy, the resulting response functions, when modelled correctly, are near perfectly correlated with the numerical bias of the underlying quantity representation.

These findings raise the possibility that the log-linear pattern displayed by young children is the result of a naïve direct response strategy. That is, very young children may not have the mensuration skills to equate a line length with quantity in the bounded number-line task (see Cohen & Sarnecka, 2014). Nevertheless, these young children may still understand the ordinal and interval properties of numbers. In this instance, the children may arbitrarily choose a certain length to correspond to a single unit. The young child may then use this line length to complete the task as we describe in the direct response strategy. In this scenario, the children's responses would necessarily result in a log-linear response function regardless of the children's underlying psychological representation of quantity.

It is, of course, possible to generate a specific quantity representation or participant strategy in which our conclusions do not hold¹. For example, a quantity representation that is both noise free and somehow scaled to the dimensions of the number-line will be mirrored in a participant's responses with a version of the direct response strategy. The ability to generate such an example, however, does not weaken our conclusions as they currently stand. The strength of a scientific model is typically gauged relative to the number of alternative hypotheses that can also explain the data. Here, we demonstrate that when the log-linear function fits well with the number-line data, this result alone is consistent with innumerable alternative models that explain that fit (e.g., different underlying quantity representations and

strategies). As a result, the log-linear fit provides no substantive evidence about the underlying psychological representation of quantities. This result reflects a fatal flaw in the many previous studies that have used such fits to support particular theoretical accounts of numerical competency. In contrast, when the CPM function fits the number-line data, currently the only model that explains this fit is the proportion/subtraction strategy applied on a specific underlying quantity representation. Until researchers propose viable alternative models that predict the CPM fits as well as the proportion/subtraction strategy model, the most plausible conclusion is that the CPM provides important information about the underlying psychological representation of the quantity.

In sum, we assert, on the basis of our computer simulations, that the bounded number-line task does not provide a transparent window on the mental representation of quantities. The constraints of the task will produce a log-linear response function when the direct response strategy is applied regardless of the underlying psychological representation of quantity. We therefore urge that researchers stop offering psychological accounts of the log-linear response functions (and hence the logarithmic, linear, or mixture of the two response functions) in the bounded number-line task that are based on particular assumptions about mental representations of quantities. We temper this negative conclusion with a more positive encouragement: if the proportion/subtraction strategy is deployed by the participant and modelled by the researcher, then one can recover useful information about the participant's quantity representation. Furthermore, researchers may also consider using the unbounded number line task (see Cohen & Blanc-Goldhammer, 2011). This avoids the problems we have described and, critically, the associated data are uncontaminated by the effects of truncation that are evident in those from the bounded number line task.

References

- Anobile, G., Cicchini, G. M., & Burr, D. C. (2012). Linear mapping of numbers onto space requires attention. *Cognition*, *122*, 454-459. doi: <http://dx.doi.org/10.1016/j.cognition.2011.11.006>
- Barth, H. C., & Paladino, A. M. (2011). The development of numerical estimation: Evidence against a representational shift. *Developmental Science*, *14*, 545-551. doi: <http://dx.doi.org/10.1111/j.1467-7687.2010.00962.x>
- Barth, H. C., Slusser, E. B., Cohen, D., & Paladino, A. M. (2011). A sense of proportion: commentary on Opfer, Siegler and Young. *Developmental Science*, *14*, 1205-1206. doi: <http://dx.doi.org/10.1111/j.1467-7687.2011.01081.x>
- Berteletti, I., Lucangeli, D., Piazza, M., Dehaene, S., & Zorzi, M. (2010). Numerical estimation in preschoolers. *Developmental Psychology*, *46*, 545-551. doi: <http://dx.doi.org/10.1037/a0017887>
- Booth, J. L., & Siegler, R. S. (2006). Developmental and individual differences in pure number estimation. *Developmental Psychology*, *41*, 189-201. doi: <http://dx.doi.org/10.1037/0012-1649.41.6.189>
- Chesney, D. L., & Matthews, P. G. (2013). Knowledge on the line: Manipulating beliefs about the magnitudes of symbolic affects the linearity of line estimation tasks. *Psychonomic Bulletin & Review*, *20*, 1146-1153. doi: <http://dx.doi.org/10.3758/s13423-013-0446-8>
- Cicchini, G. M., Anobile, G., & Burr, D. C. (2014). Compressive mapping of number to space reflects dynamic encoding mechanisms, not static logarithmic transform. *Proceedings of the National Academy of Science of the USA*, *111*, 7867-7872. doi: <http://dx.doi.org/10.1073/pnas.1402785111>
- Cohen, D. J., & Blanc-Goldhammer, D. (2011). Numerical bias in bounded and unbounded number line tasks. *Psychonomic Bulletin and Review*, *18*, 331-338. doi: <http://dx.doi.org/10.3758/s13423-011-0059-z>

- Cohen, D. J., & Sarnecka, B. (2014). Children's number-line estimation shows development of measurement skills (not number representations). *Developmental Psychology*, *50*, 1640-1652. doi: <http://dx.doi.org/10.1037/a0035901>
- Common Core State Standards Initiative. (2010). Retrieved from <http://www.corestandards.org/Math/> website:
- Dehaene, S., Izard, V., Spelke, E., & Pica, P. (2008). Log or linear? Distinct intuitions of the number scale in Western and Amazonian indigene cultures. *Science*, *320*, 1217-1220. doi: <http://dx.doi.org/10.1126/science.1156540>
- Friso-van den Bos, I., Kroesbergen, E. H., Van Luit, J. E. H., Xenidou-Dervou, I., Jonkman, L. M., Van der Schoot, M., & Van Lieshout, E. C. D. M. (2015). Longitudinal development of number line estimation and mathematics performance in primary school children. *Journal of Experimental Child Psychology*, *134*, 12-29. doi: <http://dx.doi.org/10.1016/j.jecp.2015.02.002>
- Geary, D., Hoard, M. K., Nugent, L., & Byrd-Craven, J. (2008). Development of number line representations in children with mathematical learning disability. *Developmental Neurobiology*, *33*, 277-299. doi: <http://dx.doi.org/10.1080/87565640801982361>
- Hollands, J. G., & Dyre, B. P. (2000). Bias in proportion judgments: The cyclical power model. *Psychological Review*, *107*, 500-524.
- Hurst, M., & Cordes, S. (2015). Rational Number Comparisons Across Notation: Fractions, Decimals, and Whole Numbers. *Journal of Experimental Psychology: Human Perception and Performance*, *42*, 281-293. doi: <http://dx.doi.org/10.1037/xhp0000140>
- Hurst, M., Relander, C., & Cordes, S. (2016). Biases and Benefits of Number Lines and Pie Charts in Proportion Representation. In A. Papafragou, D. Grodner, D. Mirman & J. C. Trueswell (Eds.), *Proceedings of the 38th Annual Conference of the Cognitive Science Society*. Philadelphia, PA: Cognitive Science Society.

- Laski, E. V., & Siegler, R. S. (2007). Is 27 a Big Number? Correlational and Causal Connections Among Numerical Categorization, Number Line Estimation, and Numerical Magnitude Comparison. *Child Development, 78*, 1723-1743. doi: <http://www.jstor.org/stable/4620734>
- Lu, Z., & Doshier, B. A. (2013). *Visual psychophysics: From laboratory to theory*. Cambridge, MA: MIT Press.
- Opfer, J. E., & Siegler, R. S. (2005). *Microgenetic changes in representations of numerical magnitude*. Paper presented at the Proceedings of the XXVII Annual Conference of the Cognitive Science Society, Stresa, Italy.
- Opfer, J. E., & Siegler, R. S. (2007). Representational change and children's numerical estimation. *Cognitive Psychology, 55*, 169-195. doi: <http://dx.doi.org/10.1016/j.cogpsych.2006.09.002>
- Opfer, J. E., & Siegler, R. S. (2012). Development of quantitative thinking. In K. J. Holyoak & R. G. Morrison (Eds.), *Oxford handbook of thinking and reasoning*. Oxford: UK: OUP.
- Opfer, J. E., Thompson, C. A., & Kim, D. (2016). Free versus anchored numerical estimation: A unified approach. *Cognition, 149*, 11-17. doi: <http://dx.doi.org/10.1016/j.cognition.2015.11.015>
- Siegler, R. S., & Booth, J. L. (2004). Development of numerical estimation in young children. *Child Development, 75*, 428-444. doi: <http://dx.doi.org/10.1111/j.1467-8624.2004.00684.x>
- Siegler, R. S., & Opfer, J. E. (2003). The development of numerical estimation: Evidence for multiple representations of numerical quantity. *Psychological Science, 14*, 237-243. doi: <http://dx.doi.org/10.1111/1467-9280.02438>
- Siegler, R. S., & Ramani, G. B. (2006). Early development of estimation skills. *APS Observer, 19*, 34-44.
- Siegler, R. S., Thompson, C. A., & Opfer, J. E. (2009). The Logarithmic-To-Linear Shift: One Learning Sequence, Many Tasks, Many Time Scales. *Mind, Brain*

and Education, 3, 143-150. doi: <http://dx.doi.org/10.1111/j.1751-228X.2009.01064.x>

Slusser, E. B., Santiago, R. T., & Barth, H. C. (2013). Developmental change in numerical estimation. *Journal of Experimental Psychology: General*, 142, 193-208. doi: <http://dx.doi.org/10.1037/a0028560>

Thompson, C. A., & Opfer, J. E. (2008). Costs and benefits of representational change: Effects of context on age and sex differences in symbolic magnitude estimation. *Journal of Experimental Child Psychology*, 101, 20-51. doi: <http://dx.doi.org/10.1016/j.jecp.2008.02.003>

Table 1.

Fit statistics for the log-linear model to data simulated using the direct estimate strategy with underlying quantity representations having numerical biases ranging from 0.6 (negatively accelerating) to 1.4 (positively accelerating) and varying estimated unit sizes

		Underlying Psychological Representation Distributed Gamma				
		Numerical Bias				
Unit Size		0.6	0.8	1	1.2	1.4
0.5	p(log)	0.52	0.25	0.28	0.61	0.87
	r ²	0.83	0.93	0.95	0.92	0.91
1	p(log)	0.46	0.21	0.22	0.60	0.95
	r ²	0.87	0.94	0.95	0.90	0.91
2	p(log)	0.41	0.29	0.20	0.59	0.91
	r ²	0.89	0.94	0.96	0.91	0.91
3	p(log)	0.55	0.24	0.25	0.57	0.91
	r ²	0.88	0.96	0.94	0.92	0.91
		Underlying Psychological Representation Distributed Normal with Scalar Variability				
		Numerical Bias				
Unit Size		0.6	0.8	1	1.2	1.4
0.5	p(log)	0.44	0.19	0.02	0.0	0.43
	r ²	0.98	0.98	0.99	0.98	0.88
1	p(log)	0.46	0.21	0.12	0.61	0.97
	r ²	0.98	0.98	0.99	0.89	0.88
2	p(log)	0.49	0.27	0.81	1.0	1.0
	r ²	0.98	0.99	0.90	0.88	0.88
3	p(log)	0.47	0.7	1.0	1.0	1.0
	r ²	0.98	0.93	0.84	0.84	0.80
		Underlying Psychological Representation Distributed Normal with Fixed Variability				
		Numerical Bias				
Unit Size		0.6	0.8	1	1.2	1.4
0.5	p(log)	0.47	0.20	0.02	0.0	0.32
	r ²	0.99	0.99	0.99	0.99	0.93
1	p(log)	0.49	0.21	0.02	0.45	0.86
	r ²	0.99	0.99	0.99	0.93	0.92
2						

3	p(log)	0.48	0.19	0.60	1.0	1.0
	r^2	0.99	0.99	0.93	0.92	0.75
	p(log)	0.47	0.51	1.0	1.0	1.0
	r^2	0.99	0.96	0.92	0.88	0.76

Table 2

Fit statistics for the log-linear model, one-cycle, and two-cycle CPM model to data simulated using the proportion/subtraction strategy with underlying quantity representations having numerical biases ranging from 0.6 (negatively accelerating) to 1.4 (positively accelerating)

		Underlying Psychological Representation Distributed Gamma				
		Numerical Bias				
Model		0.6	0.8	1	1.2	1.4
Log-Linear						
	p(log)	0.41	0.21	0.02	0.0	0.0
	r ²	0.95	0.98	0.99	.99	0.98
1-Cycle CPM						
	Estimated Bias	0.57	0.73	0.96	1.11	1.32
	r ²	0.96	0.98	0.99	0.99	0.99
2-Cycle CPM						
	Estimated Bias	0.54	0.73	0.87	1.10	1.32
	r ²	.99	.99	.99	.99	.99
		Underlying Psychological Representation Distributed Normal with Scalar Variability				
		Numerical Bias				
Model		0.6	0.8	1	1.2	1.4
Log-Linear						
	p(log)	0.36	0.14	0.02	0.0	0.0
	r ²	0.98	0.99	0.99	0.99	0.98
1-Cycle CPM						
	Estimated Bias	0.60	0.80	0.98	1.18	1.35
	r ²	0.99	0.99	0.99	0.99	0.99
2-Cycle CPM						
	Estimated Bias	0.57	0.77	0.98	0.16	1.37
	r ²	0.99	0.99	0.99	0.99	0.99
		Underlying Psychological Representation Distributed Normal with Fixed Variability				
		Numerical Bias				
Model		0.6	0.8	1	1.2	1.4
Log-Linear						
	p(log)	0.35	0.15	0.01	0.0	0.0
	r ²	0.98	0.99	0.99	0.99	0.98
1-Cycle CPM						
	Estimated Bias	0.61	0.80	1.00	1.19	1.37
	r ²	0.99	0.99	0.99	0.99	0.99
2-Cycle CPM						
	Estimated Bias	0.61	0.81	1.00	1.20	1.31
	r ²	0.99	0.99	0.99	0.99	0.99

Numerical Bias

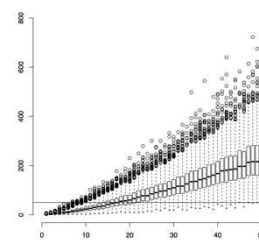
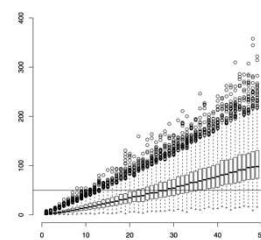
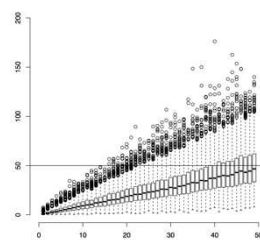
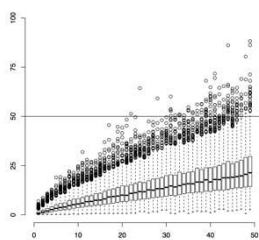
0.8

1

1.2

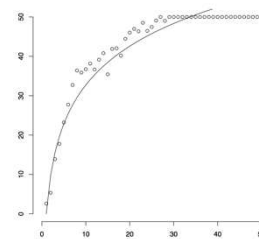
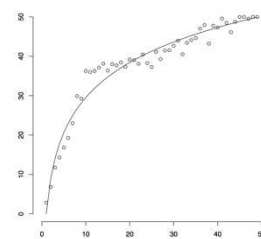
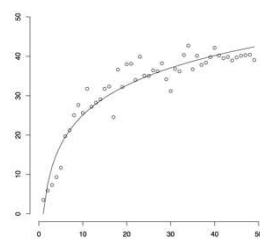
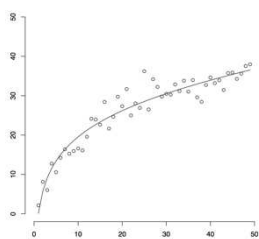
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Quantity Representation

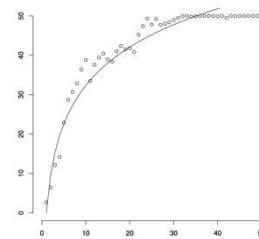
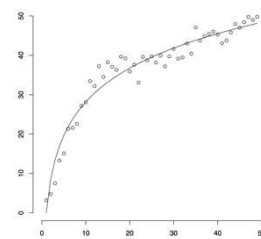
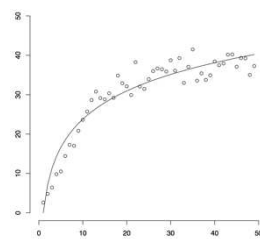
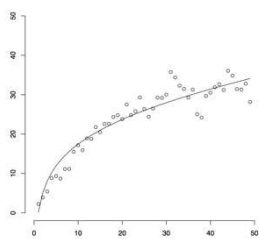


Unit Size

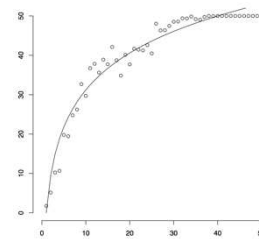
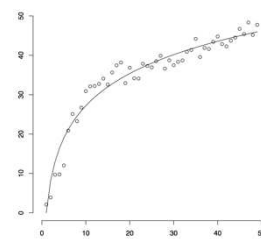
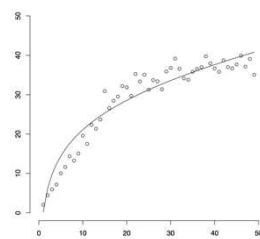
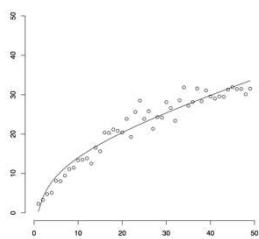
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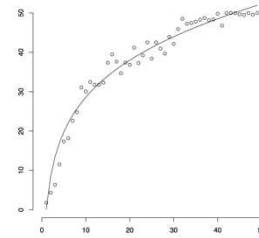
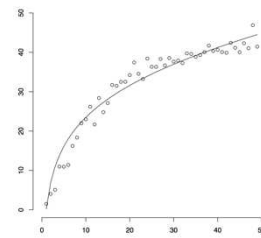
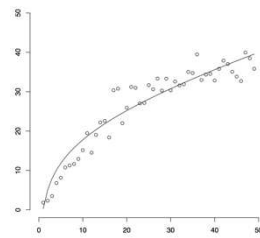
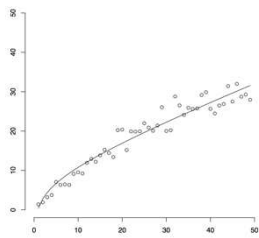
2.5



2.0



1.5



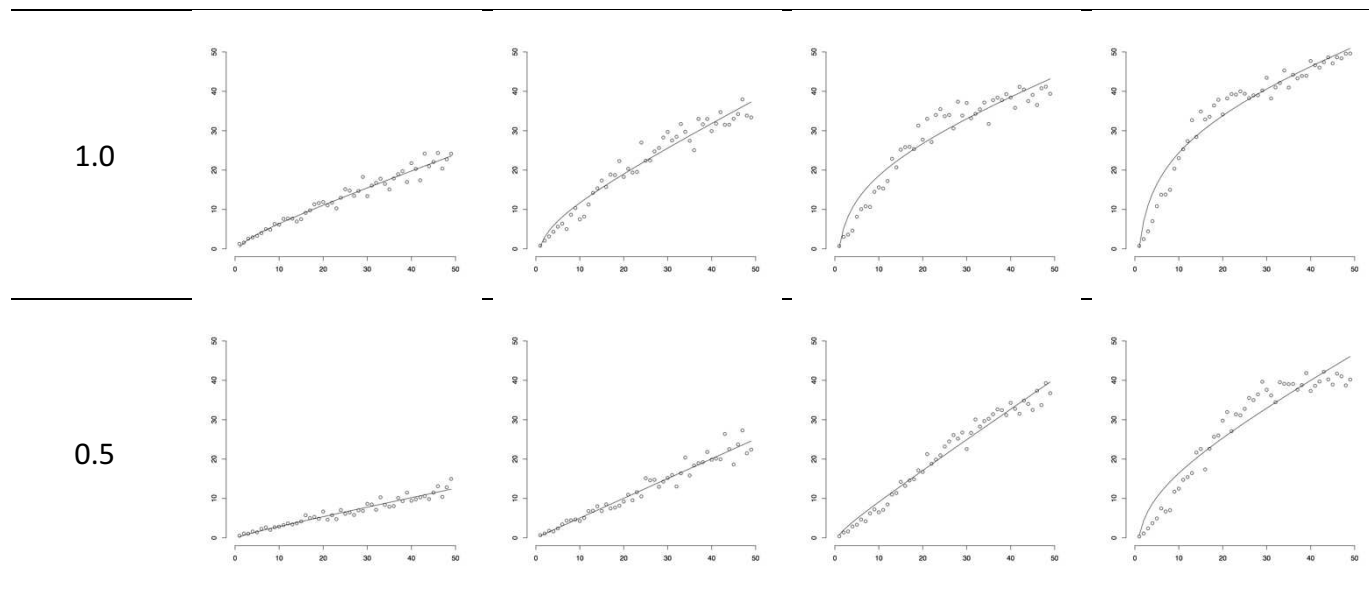


Figure 1. The simulated underlying gamma quantity representations, their resulting response estimates, and the log-linear fit to the simulated data for the direct response strategy in the bounded number-line task. Columns represent changes in numerical bias (ranging from 0.8-1.4 in 0.2 intervals) and the rows represent changes in the estimated length line corresponding to a single unit: Unit size = 1 is an accurate estimate, Unit size < 1 is an underestimate, Unit size > 1 is an overestimate.

Numerical Bias

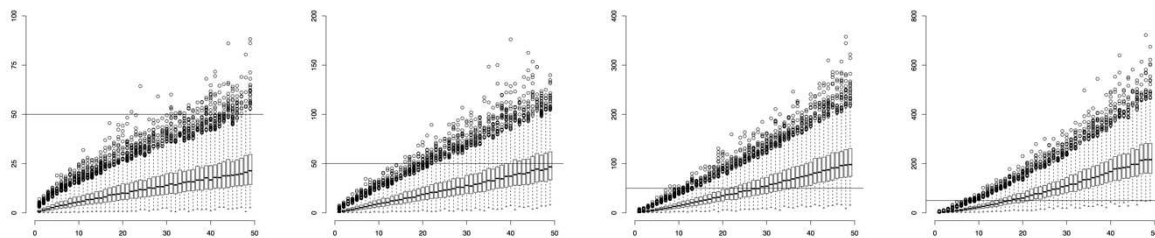
0.8

1

1.2

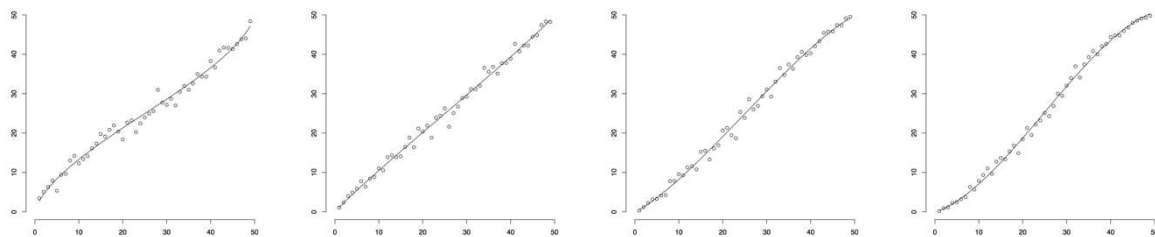
1.4

Quantity
Representation

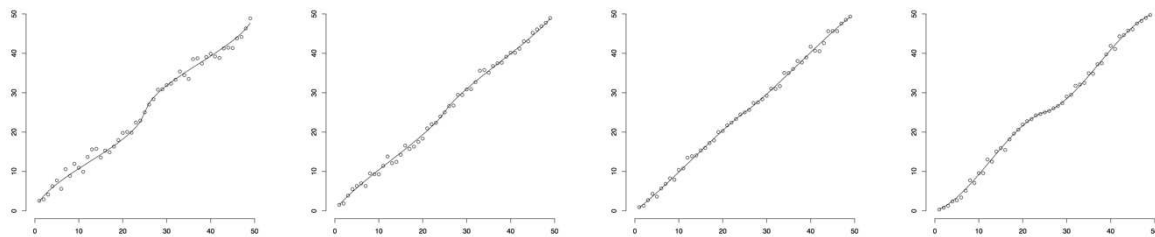


Means

1-Cycle

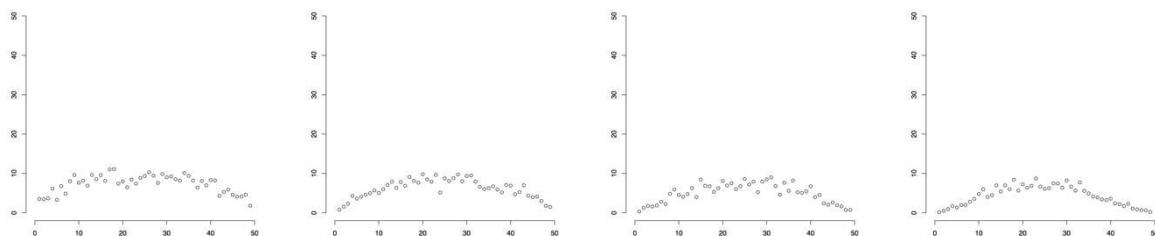


2-Cycle



SDs

1-Cycle
SDs



2-Cycle
SDs

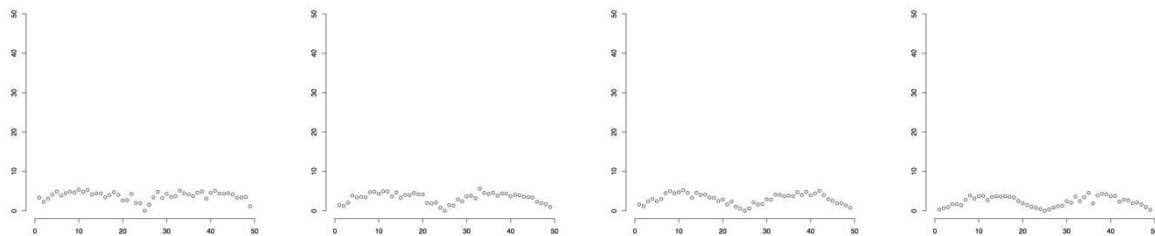


Figure 2. The simulated underlying gamma quantity representations, their resulting response estimates and SDs, and the one and two-cycle fit to the simulated data for the proportion/subtraction strategy in the bounded number-line task. Columns represent changes in numerical bias (ranging from 0.8-1.4 in 0.2 intervals)

Footnotes

¹ One may question our simulation of negatively accelerating quantity representations using power functions, rather than a log function. Here, we demonstrate that the log-linear function fits the responses derived from all quantity representations, regardless of their form (positively and negatively accelerating). Demonstrating that the log-linear function also fits the responses manifesting from a true logarithmic quantity representation is unnecessary.