¹ Rutgers University

Author Note

- Correspondence concerning this article should be addressed to Jennifer Curi. E-mail:
- jdc294@rutgers.edu

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Abstract

9 This study looks at how people compare proportions shown in different ways, like dots or

blobs. We checked if these formats change how accurate or fast people are. The results

show that some formats make it easier to compare proportions, and taking more time can

improve accuracy. These findings help us understand how the way information is shown

13 affects how people think.

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14 Keywords: keywords

Word count: X

How do different visual formats affect accuracy and reaction time in proportional reasoning tasks?

18 Introduction

Comparing proportions is sometimes very hard! But, even infants seem to be able to
do it a little bit. The purpose of this science project was better understand how well people
compare proportions when the proportions are presented in different formats. The purpose
of this class assignment is to take the R-code and plots we've been generation over the last
several weeks and put it all together into poster formal.

24 Research Objectives

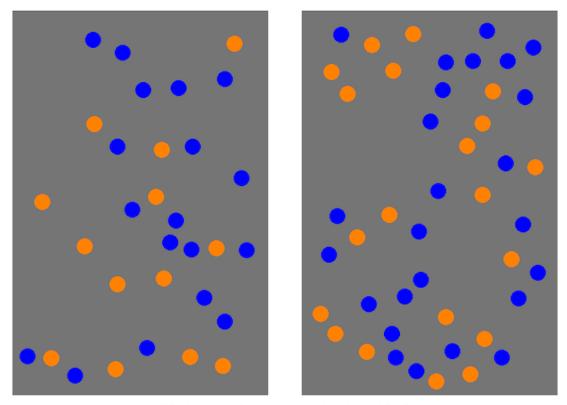
- 1. Does average performance vary across format type?
- 2. Does average performance vary across numerator congruency status?
 - 3. Does numerator congruency vary across format type? (i.e., is there an interaction)

28 Methods

9 Participants

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- A total of 99 adults participated in the study.
- First, participants were introduced to a story about a magic ball and that the outcome (i.e, blue or orange) depended on the proportions. They were then asked to compare the proportions of different images.
- In other words, participants were shown two images of the same kind at the same time and asked to decide which had a higher proportion of the shape (or dots) colored in blue.



Which has the larger proportion that is blue?

Figure 1. Results of visual stimuli used in the experiment.

Conditions

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- There were four different conditions that changed what kinds of images the 37 participants saw:
- divided blobs: blue and orange were entirely separate 39
- integrated blob: one blob, divided to be part blue and part orange 40
- separated dots: blue and orange dots were on opposite sides of the image 41
 - integrated dots: blue and orange dots were intermixed

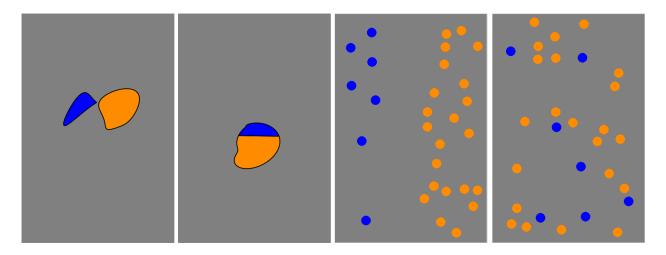


Figure 2. Types of conditions.

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43 Results

1. Does average performance vary across format type, ignoring all other aspects of the stimuli?



Figure 3. The plot shows the average performance for each condition with red dots, while grey dots show how individual responses are distributed around these averages. The average accuracy (proportion correct), varies slightly across different format conditions

2. How are reaction time and accuracy related?

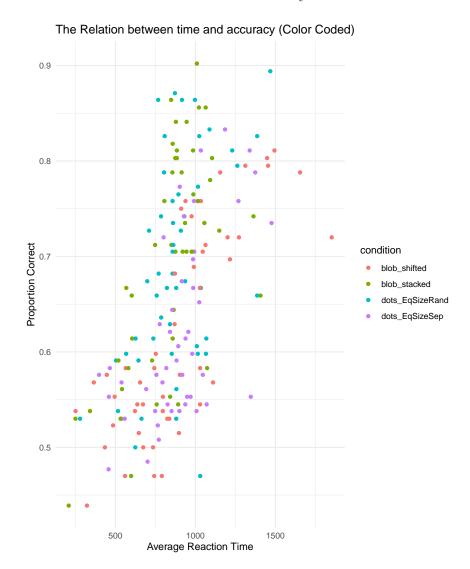


Figure 4. The plot shows that people tend to be more accurate when they take longer to respond, with different colors showing the results for each condition.

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3. How does numerator congruency interact with format type?

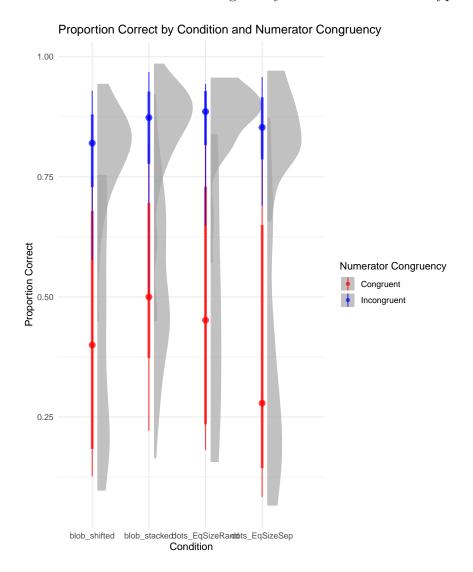


Figure 5. The plot shows that people tend to be less accurate on congruent trials (red) compared to incongruent trials (blue), and the performance varies across different conditions.

48 Discussion

49 Interpretation

Average Performance across format type: The red dots show the average
performance for each condition, while the grey dots show how individual responses vary.
Some conditions have responses that are close to the average, while others show more
spread, meaning participants' performance was less consistent.

The Correlation between reaction time and accuracy: This plot shows that
people are usually more accurate when they take longer to respond. The different colors
represent different conditions, and while the trend generally shows better accuracy with
more time, there is still some variation between the conditions.

How numerator congruency interacts with format types: The plot compares
performance on red (congruent) and blue (incongruent) trials. People tend to be less
accurate on red trials compared to blue ones. The shaded areas show how much responses
vary within each condition, with some conditions showing more spread than others.

62 Conclusion

- The tedious part of the assignment was dealing with the formatting and getting
 everything to look right on the poster. It took a lot of time to make sure each section was
 in the correct place and that the plots were accurate and displayed clearly. Because this
 was a new file, I had to re-summarize the prodtask data set and then graph it, which
 introduced some errors.
- The most satisfying part was having a template to guide me, so I didn't have to start from 0. It saved me time and made it easier to focus on putting the content together. It also felt great to see all my work come together in the end. Seeing the finished poster with everything in place was rewarding and gave me a sense of accomplishment.

Data analysis

- We used R (Version 4.3.0; R Core Team, 2023) and the R-packages dplyr (Version
- 1.1.4; Wickham, François, Henry, Müller, & Vaughan, 2023), forcats (Version 1.0.0;
- Wickham, 2023a), ggdist (Version 3.3.2; Kay, 2024), ggplot2 (Version 3.5.1; Wickham,
- 76 2016), lubridate (Version 1.9.3; Grolemund & Wickham, 2011), papaja (Version 0.1.3; Aust
- ⁷⁷ & Barth, 2024), purrr (Version 1.0.2; Wickham & Henry, 2023), readr (Version 2.1.5;
- Wickham, Hester, & Bryan, 2024), stringr (Version 1.5.1; Wickham, 2023b), tibble (Version
- 3.2.1; Müller & Wickham, 2023), tidyr (Version 1.3.1; Wickham, Vaughan, & Girlich,
- 2024), tidyverse (Version 2.0.0; Wickham, 2023c) and tinylabels (Version 0.2.4; Barth,
- 81 2023) for all our analyses.

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