

## Generalized Quantifiers and Working Memory: Disentangling Encoding from Verification

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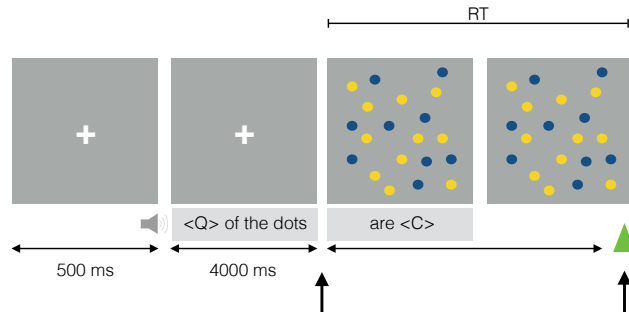
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**Overview** This pilot study employed recordings of pupil size variation in an auditory/visual verification task, to better understand working memory demands in the processing of quantified sentences. A large amount of literature has shown that the type of quantifier in a sentence significantly affects the verification procedure used to arrive at a truth-judgment [4, a.o.]. Interestingly, few studies have explored effects of quantifier type on cognitive load during early comprehension, in order to distinguish between quantifier characterization and verification procedures. We selected quantifiers from four different categories (*Aristotelian*, *Proportional*, *Numerical*, *Parity*) and exploited pupillometry measures [2, a.o.] to (I) ask whether there are effects of quantifier type on working memory specifically during *encoding*, before subjects are allowed to engage in verification; (II) if early effects on memory can be found, whether they pattern as predicted by theories of quantifier meaning grounded in the approximate number system [ANS; 1], or by computational accounts of quantifier complexity based on precise enumeration, such as the semantic automata framework [5].

**Methods** Participants ( $n = 17$ ) were asked to judge auditory sentences of the type *<Quantifier> of the dots are <Color>*, against a visual display showing systematically varied proportions of two sets of colored dots. For numerical quantifiers, the numerical referents were also varied in order to probe cardinality effects on pupil size and response time (RT). Crucially, the onset of the visual display was delayed until the onset of the disambiguating predicate, to measure increases in pupil size relative to each quantifier during *encoding* — prior to any disambiguating or search cue (i.e. the color predicate; the visual scene) — and during *verification* (Fig. 1). Each quantifier was associated to two target colors (*blue*, *yellow*) in two verification conditions (*true*, *false*). Proportions of colors in the visual arrays were varied so to avoid fixed counting strategies. Each quantifier was presented 24 times, for a total of 216 trials. SR Research DataViewer was used to output trial reports for three distinct interest periods: baseline, encoding, and verification. For each interest period, we fit linear-mixed models in R with RT, mean, and max. pupil response as dependent variables; Quantifier Type (4 levels) and Proportion (14 levels) as fixed effects, and Participant as a random effect.

**Encoding** (Fig. 2a). We found effects of quantifier type on mean ( $F(3,3190) = 7.36$ ,  $p < 0.001$ ) and max ( $F(3,3190) = 8.14$ ,  $p < 0.001$ ) pupil response during encoding, confirming that there were comprehension effects on working memory guided by the semantic content of different quantifiers. Quantifier effects clustered in two main groups: Aristotelian-Proportional (AP) quantifiers eliciting significantly smaller pupil responses than Parity-Numerical (PN), with no significant differences found within clusters. **Verification** (Fig. 2b). Significant effects were found of quantifier type on mean ( $F(3,3189) = 5.117$ ,  $p < 0.01$ ) and max ( $F(3,3190) = 31.740$ ,  $p < 0.001$ ) pupil response during verification. Again AP quantifiers showed significantly smaller pupil responses than for PN (see Figure 2), with no significant differences within AP (mean:  $p < 0.16$ ; max:  $p < 0.94$ ) and PN (mean:  $p < 0.63$ ; max:  $p < 0.55$ ) clusters, respectively. **RTs** (Fig. 2c). We found effects on response times both for quantifier category ( $F(3,3189) = 662.23$ ,  $p < 0.001$ ) and proportion ( $F(15,3189) = 11.37$ ,  $p < 0.001$ ), with RTs faster for Aristotelian < Proportional < Parity/Numerical.

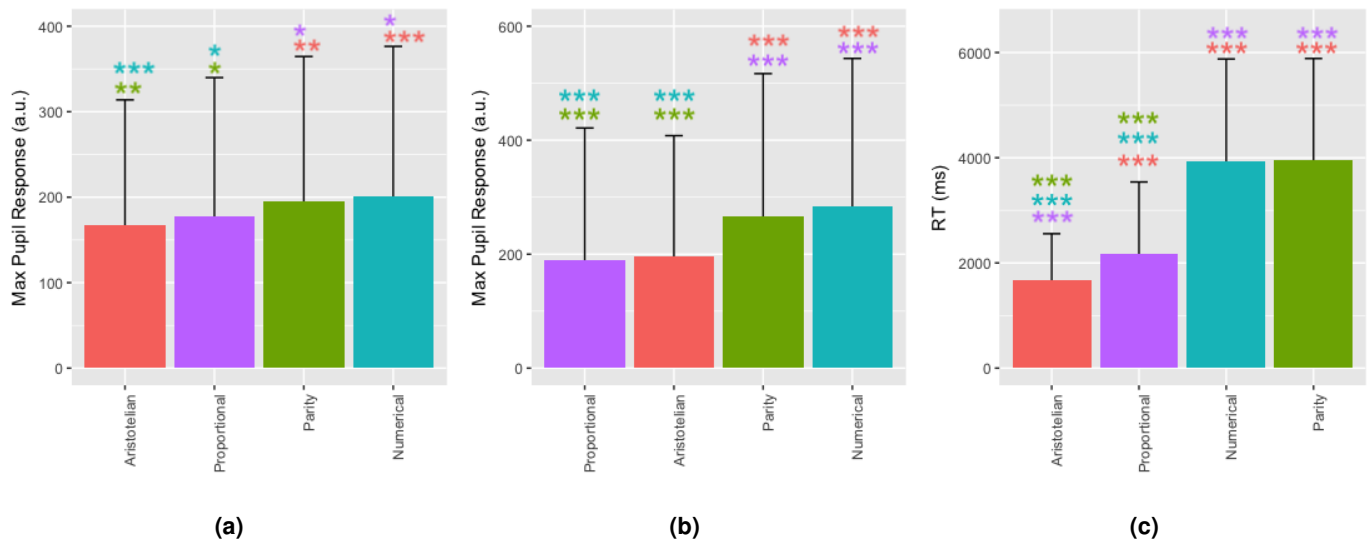
**Discussion** These preliminary results suggest that quantified expressions modulate working memory already during early comprehension, before any cue to verification has been given. Pupil effects suggest that the initial specification of proportional quantifiers relies on approximate comparisons between sets instead of precise one-to-one counting. Bigger effects recorded for Numerical and Parity quantifiers in encoding are consistent with the idea of additional working memory load required for the encoding of precise numerical concepts, and to early recruitment of cognitive resources that will later be needed by the verification procedure associated to different quantifiers [3]. Future work should probe the role of varying proportions in modulating RTs and pupil response in verification.



**Figure 1:** Experimental design.

Quantifier	Magnitude	Quantifier Category
All		Aristotelian
No		
Some		
At least $n$	$n = 2, \dots, 7; 9 \dots 14$	Numerical
At most $n$	$n = 2, \dots, 7; 9 \dots 14$	
An even number of		Parity
An odd number of		
Most		Proportional
More than half		

**Table 1:** Quantifiers grouped by category



**Figure 2:** Comparisons of means by quantifier category for max pupil response (in arbitrary units) during encoding (a) and verification (b); and for RTs (in ms) from image onset to end of trial (c). Signif. codes (\*\*\*: 0.001; \*\*: 0.01; \* : 0.05) are color coded by the quantifier category of reference.

## References

- [1] S. Dehaene. *The number sense: How the mind creates mathematics*. OUP USA, 1999.
- [2] P. E. Engelhardt, F. Ferreira, and E. G. Patsenko. Pupillometry reveals processing load during spoken language comprehension. *The Quarterly Journal of Experimental Psychology*, 63(4):639–645, 2010.
- [3] J. Lidz, P. Pietroski, J. Halberda, and T. Hunter. Interface transparency and the psychosemantics of most. *Natural Language Semantics*, 19(3):227–256, 2011. ISSN 1572-865X. doi: 10.1007/s11050-010-9062-6. URL <http://dx.doi.org/10.1007/s11050-010-9062-6>.
- [4] P. Pietroski, J. Lidz, T. Hunter, and J. Halberda. The meaning of most: Semantics, numerosity and psychology. *Mind & Language*, 24(5):554–585, 2009.
- [5] J. Szymanik. Computing simple quantifiers. In *Quantifiers and Cognition: Logical and Computational Perspectives*, pages 41–49. Springer International Publishing, 2016.