

# Scales of Negativity

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(part of this research is joint work with Karen De Clercq)

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# Introduction

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Simple answer:

- A sentence is **negative** iff it involves **sentential negation**.
- In English: *not* (possibly contracted and with *do* support)

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| 2. Sue <b>didn't</b> leave. | negative |

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  - In English: *not* (possibly contracted and with *do* support)
1. Sue left. positive
  2. Sue **didn't** leave. negative

## Standard syntactic and semantic assumptions

- Syntactically, negation forms a high functional projection (Laka 1990, Haegeman and Zanutinni 1991, Haegeman 1995)
- Semantically, negation amounts to **complementation**

# Beyond the simple case

## Negativity without overt sentential negation

Klima 1964, Ladusaw 1992, Zeijlstra 2004, Penka 2011, Tubau 2008 a.o.

- |                            |             |
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| 3. Sue <b>never</b> left.  | N-WORD-ADV  |
| 4. <b>Nobody</b> left.     | N-WORD-SUBJ |
| 5. Sue saw <b>nobody</b> . | N-WORD-OBJ  |

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| 6. Sue <b>rarely</b> left.       | DE-ADV      |
| 7. <b>Few students</b> saw Sue.  | DE-SUBJ     |
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There is a whiff of negativity in all these examples, but . . .

- are they as negative as the simple case (*Sue didn't leave*)?
- are there different levels/scales of negativity?

# Today

## Testing negativity using two criteria

### I. Polarity particle responses

(new)

- Agreeing with positive sentences  $\Rightarrow$  YES / \*NO
- Agreeing with negative sentences  $\Rightarrow$  YES / NO

### II. Question tags

(Klima 1964)

- Positive sentences  $\Rightarrow$  negative tags
- Negative sentences  $\Rightarrow$  positive tags



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## Outcome: two scales of negativity

I. **Semantic scale**: N-WORD  $\gg$  DE

II. **Syntactic scale**: ADV  $\gg$  SUBJ  $\gg$  OBJ

# Roadmap

## I. Testing negativity using **polarity particles**

- Experiment 1: 'testing the test'
- Experiment 2: applying the test

## II. Testing negativity using **question tags**

- Experiment 3

## III. Discussion: **scales of negativity**

## IV. Conclusions and future work

## Polarity particles

It has been claimed that the felicity of polarity particles in responses to assertions and polar questions is sensitive to:

- The nature of the response: **agreeing** or **disagreeing**
- The polarity of the antecedent: **positive** or **negative**

(Pope 1976, Ginzburg & Sag 2000, Kramer & Rawlins 2009, Farkas 2011, Holmberg 2012, Cooper & Ginzburg 2012, Farkas & Roelofsen 2012)

## Polarity particles in agreeing responses

In an agreeing response to a **positive** assertion, **only YES** can be used:

9. A: Paul stepped forward.

B: Yes / \*No, Paul stepped forward.

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In an agreeing response to a **negative** assertion, **both YES and NO** can be used:

10. A: Paul did not step forward.  
B: Yes / No, Paul did not step forward.

# Testing negativity using polarity particles

## The simplest cases

- Agreeing with positive sentences  $\Rightarrow$  YES / \*NO
- Agreeing with negative sentences  $\Rightarrow$  YES / NO

## More generally

The **frequency of NO** in an agreeing response can be seen as an indicator of the **extent to which the antecedent is negative**

frequency of NO in agreeing responses



negativity of the antecedent

# Experiment 1: testing the test

## Question

Our test assumes that in the simplest cases:

- Agreeing with positive sentences  $\Rightarrow$  YES / \*NO
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This is a prediction of some (though not all) approaches in the theoretical literature. **But is it real?**

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## Answer

- Description of experiment.
- Discussion of results.
- Conclusion: usage of NO in agreeing responses is indeed a good test for diagnosing negativity of the antecedent.



# Experiment 1

## Example items

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{ Yes, it will. }

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12. At most six volunteers did not sign up for free housing.  
{ Yes, at most six of them didn't. }  
{ No, at most six of them didn't. }

# Experiment 1

## Response variable

**resp** encodes choice of polarity particle in responses;  
factor with 2 levels: YES ('success'); NO.

## Two predictors

**stim-pol** encodes polarity of stimulus;  
factor with 2 levels: POS (ref. level), NEG.

- If POS, we expect agreement with YES.
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We want to see if the referential vs. quantificational nature of the subject NPs and their monotonicity properties affect particle choice.

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## Example items, again

All stimuli have the structure **subject** + **predication**.

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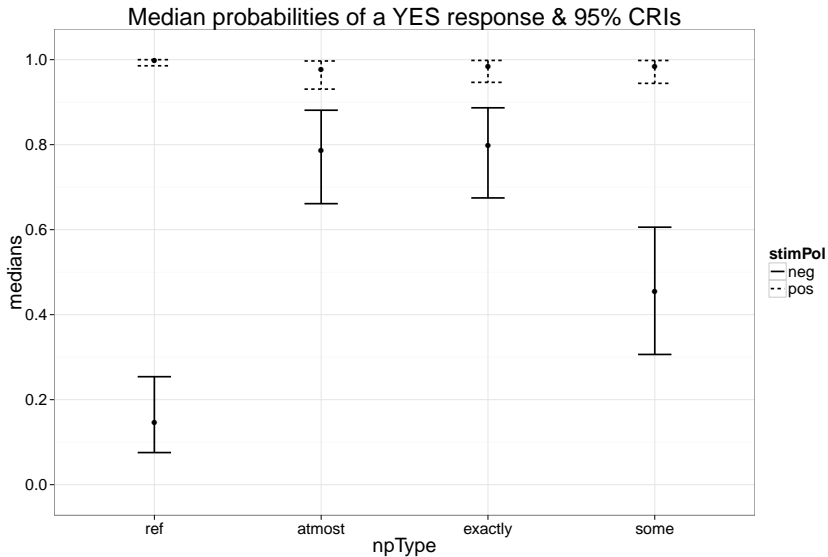
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At most six volunteers did not sign up for free housing.

{ Yes, at most six of them didn't. }  
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- I. POS stimuli with REF subjects license **only YES**.
- II. NEG stimuli with REF subjects license **both YES and NO**, as expected, with a rather **strong preference for NO**.
- III. POS sentences: changing NP type diminishes preference for YES compared to REF subjects; decrease significant only for downward-entailing ATMOST (see appendix).
- IV. NEG sentences: **non-REF subjects neutralize or even reverse the preference** for NO found with REF NPs.
- V. This interaction of NEG polarity and NP type is not predicted by the theoretical literature to date.

# Experiment 1: Conclusion

Main result: prediction confirmed

- The distribution of YES and NO in agreeing responses is indeed **sensitive to the polarity of the antecedent**.
- **Negative** antecedents license **NO** in agreeing responses; **positive** antecedents don't.

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## Additional results

- **Preference for NO** over YES in agreeing responses to negative assertions with **REF** subjects.  
(predicted by the account of Farkas & Roelofsen 2012)
- This preference is **neutralized** by existential subjects and even **reversed** by downward/non-monotonic subjects.  
(not predicted by any existing theoretical account)

## Experiment 2: Applying the test

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### Eight conditions

- Six 'treatment' conditions:

**two semantic types** × **three syntactic positions**  
(N-WORD, DE)                      (ADV, SUBJ, OBJ)

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### Eight conditions

- Six 'treatment' conditions:

**two semantic types** × **three syntactic positions**  
(N-WORD, DE)                      (ADV, SUBJ, OBJ)

- Two **control** conditions: **POSITIVE** and **NEGATIVE** sentences in which the adverb, subject and object are referential.

## Experiment 2: Stimuli

- All stimuli have the following structure:

subj + verb + obj + adv

- Example:

The representatives + visited + the colonies + this year.

- The arguments and the adverb are always **referential** except when experimentally manipulated.
- When **referential**, the adverb is more natural **after object**; when **manipulated**, it is more natural **before verb**.

## Experiment 2: Example items

### Controls

13. **cond** = POSITIVE

The representatives visited the colonies this year.

{ Yes, they did. }  
{ No, they did. }

14. **cond** = NEGATIVE

The representatives didn't visit the colonies this year.

{ Yes, they didn't. }  
{ No, they didn't. }

## Experiment 2: Example items

### N-WORDS

15. **cond** = N-WORD-ADV

The representatives never visited the colonies.

{ Yes, they never did. }  
{ No, they never did. }

16. **cond** = N-WORD-SUBJ

No representatives visited the colonies this year.

{ Yes, no representatives did. }  
{ No, no representatives did. }

17. **cond** = N-WORD-OBJ

The representatives visited no colonies this year.

{ Yes, they visited no colonies. }  
{ No, they visited no colonies. }

## Experiment 2: Example items

### DE-quantifiers

18. **cond** = DE-ADV

The representatives rarely visited the colonies.

{ Yes, they rarely did. }  
{ No, they rarely did. }

19. **cond** = DE-SUBJ

Few representatives visited the colonies this year.

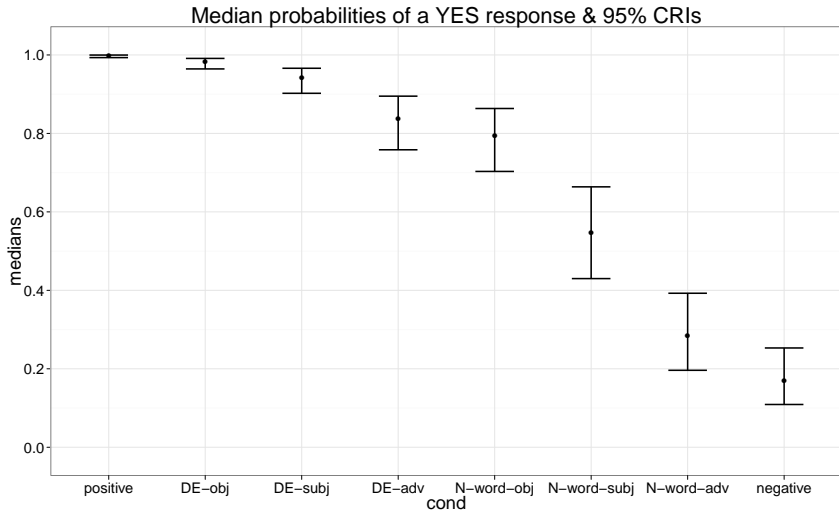
{ Yes, few representatives did. }  
{ No, few representatives did. }

20. **cond** = DE-OBJ

The representatives visited few colonies this year.

{ Yes, they visited few colonies. }  
{ No, they visited few colonies. }

# Experiment 2: Results





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## Generalizations

- Once again, we see that the POSITIVE and NEGATIVE (and across-the-board referential) **controls behave as expected**:
  - probability of YES for POSITIVE is practically 1
  - probability of YES for NEGATIVE is lowest, about 0.17
- The **biggest split** is the **semantic** one:  
N-WORDS are clearly more negative than DE-quantifiers.  
(although N-OBJ is not significantly more negative than DE-ADV)
- **Within each semantic class**, we have a clear, statistically significant **syntactic hierarchy**: ADV is more negative than SUBJ, which in turn is more negative than OBJ.

## Experiment 2: Discussion

### Scales of negativity

We have two scales of negativity, a semantic one and a syntactic one:

**Semantic scale of negativity:** N-WORD  $\gg$  DE

**Syntactic scale of negativity:** ADV  $\gg$  SUBJ  $\gg$  OBJ

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The semantic scale is primary: (unsurprisingly) semantics takes precedence in determining the negativity of a sentence.

Each of the two semantically-ordered strata is further refined by the syntactic scale (more surprising).

## Experiment 2: Discussion

### The semantic scale and NPI licensing

The semantic scale is reminiscent of the NPI literature

(Zwarts 1995, Levinson 2008, Giannakidou 2011, Hoeksema 2012, a.o.)

- **Weak** NPIs are licensed by **downward entailing** elements
- **Strong** NPIs are only licensed by **anti-additive** elements
- Anti-additive elements are always downward entailing, but not vice versa.
- N-WORDS are **anti-additive**.
- DE quantifiers are **downward entailing**, not anti-additive.
- So wrt NPI licensing we have the **same semantic scale**:

N-WORD  $\gg$  DE

## Experiment 2: Discussion

### Downward entailingness and anti-additivity

A function  $\mathcal{F}(\cdot)$  of quantifier type  $\langle\langle et \rangle t \rangle$  is:

- **DOWNWARD ENTAILING** iff  $\mathcal{F}(Y) \rightarrow \mathcal{F}(X)$  for any  $X \subseteq Y$ .
- *Few students are tall*  $\rightarrow$  *Few students are tall and blond*
- **ANTI-ADDITIVE** iff  $\mathcal{F}(X \cup Y) \leftrightarrow \mathcal{F}(X) \wedge \mathcal{F}(Y)$  for any  $X, Y$
- *No students are tall or blond*  $\leftrightarrow$   
*No students are tall*  $\wedge$  *No students are blond*

Every anti-additive function is also downward entailing,  
but **not every downward entailing function is also anti-additive.**

- *Few students are tall or blond*  $\leftrightarrow$   
*Few students are tall*  $\wedge$  *Few students are blond*

## Experiment 2: Discussion

### N-words versus other anti-additive elements

- Broadly speaking, the semantic scales relevant for polarity particles and NPI licensing align very well.
- However, there also seems to be a **subtle difference**.
- There seems to be something special about N-WORDS, besides their general anti-additive nature, that determines their behavior wrt polarity particles.
- For example, *without* is also anti-additive, but seems to behave differently (still to be confirmed experimentally):

21. Bill ate **without** a spoon.

Yes / **\*No**, he ate without a spoon.

## Coming next: Experiment 3

Checking whether the existence of the two scales of negativity is independently confirmed by another, more traditional test.

## Experiment 3: Question tags

### Same main question

Do sentence with N-WORDS and DE quantifiers behave like **negative** or like **positive** sentences?

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Hence, **2 controls** and **6 treatments (cond)** just as in Exp. 2  
× **2 contexts (conx)**, for a total of **16 conditions**:

- **REV contexts**: interlocutor is an authority for content of assertion, so q-tag used to ask for confirmation; q-tag polarity is the **REVERSE** of sentence polarity.

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- **REV contexts**: interlocutor is an authority for content of assertion, so q-tag used to ask for confirmation; q-tag polarity is the **REVERSE** of sentence polarity.
- **RED contexts**: interlocutor finds content of assertion hard to believe, so q-tag used to express skepticism / suspicion / sarcasm; q-tag polarity **REDUPLICATES** sentence polarity.

## Experiment 3: Example items

Items have same basic structure as in Exp. 2.

### Controls

23. **cond** = POSITIVE, **conx** = REV

Mary has just told Jane something about government representatives. Jane **knows it already** and says:  
“**Aha, I was right!** The government representative visited the colonies this year, { did he? }  
{ didn't he? }”

**cond** = POSITIVE, **conx** = RED

Mary has just told Jane something about government representatives. Jane **finds it hard to believe** and says:  
“**You don't say!** The government representative visited the colonies this year, { did he? }  
{ didn't he? }”

## Experiment 3: Example items

### Controls (ctd.)

24. **cond** = NEGATIVE, **conx** = REV

Geoff has just told Ryan something about composers.

Ryan knows it already and says:

“Aha, I was right! The composer did **not** use the cello in his late period, { did he? }”

**cond** = NEGATIVE, **conx** = RED

Geoff has just told Ryan something about composers.

Ryan finds it hard to believe and says:

“You don’t say! The composer did **not** use the cello in his late period, { did he? }”

## Experiment 3: Example items

### N-WORDS

25. **cond** = N-WORD-ADV, **conx** = REV

Lilly has just told Adrian something about the children.  
Adrian knows it already and says:

“Aha, I was right! The children **never** understood the  
science experiments, { did they? }”  
{ didn't they? }

- cond** = N-WORD-ADV, **conx** = RED

Lilly has just told Adrian something about the children.  
Adrian finds it hard to believe and says:

You don't say! The children **never** understood the science  
experiments, { did they? }”  
{ didn't they? }

## Experiment 3: Example items

### DE-quantifiers

26. **cond** = DE-ADV, **conx** = REV

Bob has just told Justine something about the athletes.

Justine knows it already and says:

“Aha, I was right! The athletes **rarely** skipped the practice games, { did they? },”  
{ didn't they? }

- cond** = DE-ADV, **conx** = RED

Bob has just told Justine something about the athletes.

Justine finds it hard to believe and says:

You don't say! The athletes **rarely** skipped the practice games, { did they? },”  
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- **POS** sentences expected to **discriminate** very well between REV and RED contexts.

They should trigger:

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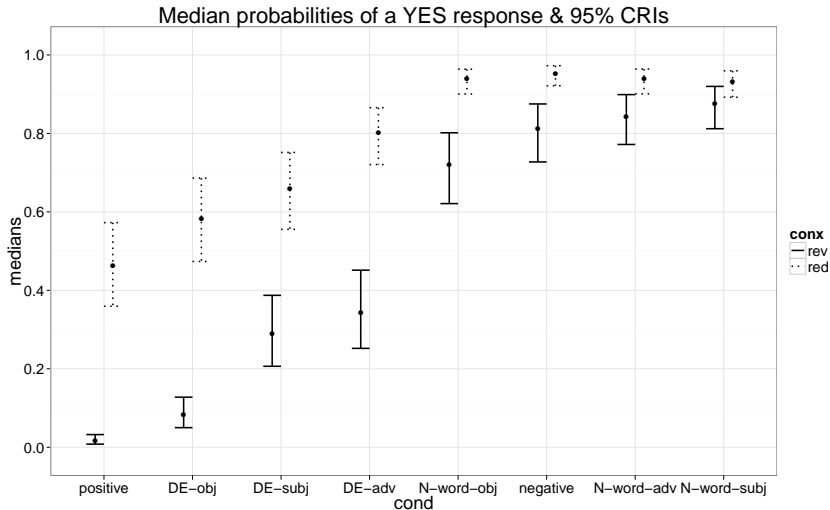
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- **POS** sentences expected to **discriminate** very well between REV and RED contexts.  
They should trigger:
  - NO (negative tags) in REV contexts, and
  - YES (positive tags) in RED contexts.
- **NEG** sentences expected to be ungrammatical with NO; they should **not discriminate** between REV and RED.
  - E.g: *Sue didn't call, \*didn't she?*

(Quirk *et al.* 1985, McCawley 1998; turns out: not true).

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- The **controls didn't work as expected in RED**; we will therefore **mainly focus on REV**.
- Both POS and NEG **controls** worked more or less as expected in REV.
- **POS** in REV clearly the most positive, hence probability of getting a YES (positive tag) response very close to 0.
- **NEG** in REV among the most negative, hence probability of getting a YES (positive tag) response close to 1.

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## Generalizations (ctd.)

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- **Semantic scale** of negativity clearly **confirmed**:  
N-WORD sentences are more negative than DE sentences.
- **Syntactic scale** of negativity **confirmed, but not completely**.  
For each of the 2 semantically ordered strata, OBJ is clearly less negative than SUBJ and ADV, but there is **no significant difference between ADV and SUBJ**.

## Experiment 3: Discussion

### Main result: scales of negativity confirmed

Experiment 3 confirms the results of experiment 2:

- There are **two scales of negativity**, one semantic and one syntactic.
- The semantic scale takes precedence and the syntactic scale further refines it.



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### Q-tags noisier than polarity particles

- **Semantic scale** completely confirmed: N-WORD  $\gg$  DE
- **Syntactic scale** coarser grained: {ADV, SUBJ}  $\gg$  OBJ

# Conclusion and future research

## Most important findings

- Negativity of a sentence (as measured by polarity particles and q-tags) is a **matter of degree**.
- There are **two interacting scales** that influence negativity.

**Semantic scale:** N-WORD  $\gg$  DE

**Syntactic scale:** ADV  $\gg$  SUBJ  $\gg$  OBJ

## Questions for future research (1)

How to formally capture the two scales and their interaction?

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How to formally capture the two scales and their interaction?

- Semantic scale maybe explained by N-WORDS being more negative than DE (anti-additive  $\Rightarrow$  de)

How about the syntactic scale?

- Most direct approach: assume that N-WORDS and DE items both have syntactic negative features, the former stronger than the latter.
- Assume that the negative strength of a sentence depends on the negative strength of the negative items occurring in that sentence, as well as their syntactic position.
- The latter dependency could perhaps be spelled out in terms of agreement.
- The negative indefinite approach to N-WORDS may help to explain the observed contrast between N-WORDS and other anti-additive items.

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- What explains the interaction of polarity particle choice and nature of quantificational NP observed in Experiment 1?

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- What explains the interaction of polarity particle choice and nature of quantificational NP observed in Experiment 1?
- Hypothesis: the more complex a negative sentence is, the higher the frequency of YES in agreeing responses.



## Questions for future research (2)

- What explains the interaction of polarity particle choice and nature of quantificational NP observed in Experiment 1?
- Hypothesis: the more complex a negative sentence is, the higher the frequency of YES in agreeing responses.
- In particular: the more complex the quantifiers in the NEG sentence, the higher the rate of YES in agreeing responses.

quantifier complexity



frequency of YES in agreeing with NEG sentences

## Questions for future research (2)

- What explains the interaction of polarity particle choice and nature of quantificational NP observed in Experiment 1?
- Hypothesis: the more complex a negative sentence is, the higher the frequency of YES in agreeing responses.
- In particular: the more complex the quantifiers in the NEG sentence, the higher the rate of YES in agreeing responses.

quantifier complexity



frequency of YES in agreeing with NEG sentences

- Hypothesis confirmed in a self-paced reading experiment:
  - I. Quantifiers are more complex than referential NPs.
  - II. Downward entailing (and non-monotonic) quantifiers are more complex than upward entailing ones.
  - III. Modified numerals are more complex than unmodified quantifiers.

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# Appendix: Experiment 1

## Data Analysis

The response variable **resp** is categorical binary, so mixed-effects logistic regression models:

- fixed effects: **stim-pol**, **np-type** and their interaction;
- random effects: correlated subject random effects for the intercept and the **stim-pol**-NEG slope.
- random effects for subjects factor out the variability in responses between subjects; whatever variability remains more confidently attributed to the experimental manipulations
- the items did not account for any variability in this experiment, so we do not include item random effects

When **stim-pol** = POS, most counts are 0 or extremely low, so usual frequentist procedures less reliable.

But: Bayesian modeling with the same likelihood structure and vague / low-information priors OK.



# Appendix: Experiment 1

## Priors for the fixed effects

Priors for the intercept and the non-reference levels of **stim-pol**, **np-type** and their interaction:

- All independent normals  $\mathcal{N}(0, 10^2)$ .
- These priors place most of their probability on the interval  $(-20, 20)$ , a very wide interval on the standard logit scale.
- Therefore, they contribute very little information and the posterior estimates overwhelmingly reflect the data.

## Priors for the random effects

Priors for the random effects are similarly vague / low information:

- A bivariate normal distribution for the intercept and **stim-pol**-NEG random effects with correlation  $\rho$  between the two random effects

$$\mathcal{N} \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \sigma^2 & \rho\sigma\tau \\ \rho\sigma\tau & \tau^2 \end{bmatrix} \right).$$

Priors for the intercept standard deviation  $\sigma$  and the **stim-pol**-NEG standard deviation  $\tau$ : independent uniforms  $\mathcal{U}(0, 10)$ .

Prior for  $\rho$ :  $\mathcal{U}(-1, 1)$ .

## Appendix: Experiment 1

Estimated means, standard deviations and 95% credible intervals (CRIs) for posterior distributions of random and fixed effects (logit scale):

RANDOM EFFECTS		mean	std.dev.	95% CRI
	$\sigma$	0.978	0.661	(0.057, 2.570)
	$\tau$	1.614	0.591	(0.568, 3.025)
	$\rho$	-0.262	0.500	(-0.906, 0.866)
FIXED EFFECTS		mean	std.dev.	95% CRI
	<b>intercept</b>	7.104	1.874	(4.228, 11.526)
	<b>stim-pol-NEG</b>	-8.868	1.894	(-13.296, -5.891)
	<b>np-type-ATMOST</b>	-3.249	1.849	(-7.561, -0.268)
	<b>np-type-EXACTLY</b>	-2.877	1.898	(-7.314, 0.218)
	<b>np-type-SOME</b>	-2.903	1.873	(-7.183, 0.201)
	<b>stim-pol-NEG</b> :	6.326	1.885	(3.206, 10.676)
	NP-TYPE- <b>atmost</b>			
	<b>stim-pol-NEG</b> :	6.015	1.933	(2.823, 10.474)
	NP-TYPE- <b>exactly</b>			
	<b>stim-pol-NEG</b> :	4.481	1.898	(1.247, 8.805)
	NP-TYPE- <b>some</b>			

MCMC: 3 chains, 275000 iterations per chain, 25000 burn-in, 100 thinning.

## Appendix: Experiment 2

### Data Analysis

The response variable **resp** is categorical binary, so mixed-effects logistic regression models:

- fixed effect: **cond** (factor with 8 levels, reference level: POSITIVE)
- random effects: intercept random effects for subjects
- the items or more complex random effects structures for subjects or items accounted for (practically) no variability in this experiment, so we do not include item random effects

## Appendix: Experiment 2

### Priors for the fixed effects

Priors for the intercept and the non-reference levels of **cond** (reference level: POSITIVE):

- All independent normals  $\mathcal{N}(\mu = 0, \sigma^2 = 1000)$ .
- These priors place most of their probability on the interval  $(-65, 65)$ , which is an extremely wide interval on the standard logit scale.
- Therefore, they contribute very little information and the posterior estimates overwhelmingly reflect the data.

### Priors for the random effects

Priors for the random effects are similarly vague / low information:

- A normal distribution for the intercept random effects for subjects  $\mathcal{N}(0, \sigma^2)$ .

Prior for the intercept standard deviation  $\sigma$ : a uniform  $\mathcal{U}(0, 100)$ .

## Appendix: Experiment 2

Estimated means, standard deviations and 95% credible intervals (CRIs) for posterior distributions of random and fixed effects (logit scale):

RANDOM EFFECTS		mean	std.dev.	95% CRI
	$\sigma$	1.656	0.182	(1.336, 2.051)
FIXED EFFECTS		mean	std.dev.	95% CRI
	INTERCEPT	6.368	0.839	(4.991, 8.242)
	<b>cond</b> -DE-SUBJ	-3.592	0.830	(-5.445, -2.221)
	<b>cond</b> -N-WORD-SUBJ	-6.174	0.822	(-8.021, -4.822)
	<b>cond</b> -DE-OBJ	-2.375	0.854	(-4.276, -0.925)
	<b>cond</b> -N-WORD-OBJ	-5.019	0.823	(-6.886, -3.675)
	<b>cond</b> -DE-ADV	-4.732	0.825	(-6.605, -3.389)
	<b>cond</b> -N-WORD-ADV	-7.291	0.828	(-9.143, -5.937)
	<b>cond</b> -NEGATIVE	-7.962	0.833	(-9.858, -6.594)

MCMC: 3 chains, 250000 iterations per chain, 25000 burn-in, 50 thinning.

## Appendix: Experiment 3

### Data Analysis

The response variable **resp** is categorical binary, so mixed-effects logistic regression models:

- fixed effects: **cond** (factor with 8 levels, reference level: POSITIVE), **conx** (factor with 2 levels, reference level: RED) and their interaction
- random effects: subject random effects for the intercept and **conx**-REV slope
- once again, the items did not account for any variability, so we do not include item random effects

## Appendix: Experiment 3

### Priors for the fixed effects

Priors for the intercept and the non-reference levels of **cond** (reference level: POSITIVE) and **conx** (reference level: RED) and their interaction:

- All independent normals  $\mathcal{N}(\mu = 0, \sigma^2 = 1000)$ .
- These priors place most of their probability on the interval  $(-65, 65)$ , which is an extremely wide interval on the standard logit scale.
- Therefore, they contribute very little information and the posterior estimates overwhelmingly reflect the data.

### Priors for the random effects

Priors for the random effects are similarly vague / low information:

- A bivariate normal distribution for the intercept and **conx**-REV random effects with correlation  $\rho$  between the two random effects

$$\mathcal{N}\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \sigma^2 & \rho\sigma\tau \\ \rho\sigma\tau & \tau^2 \end{bmatrix}\right).$$

Priors for the intercept standard deviation  $\sigma$  and the **conx**-REV standard deviation  $\tau$ : independent uniforms  $\mathcal{U}(0, 100)$ .

Prior for  $\rho$ :  $\mathcal{U}(-1, 1)$ .

## Appendix: Experiment 3

Estimated means, standard deviations and 95% credible intervals (CRIs) for posterior distributions of random and fixed effects (logit scale):

RANEFS		mean	std.dev.	95% CRI
	$\sigma$	1.794	0.169	(1.484, 2.149)
	$\tau$	2.655	0.227	(2.243, 3.143)
	$\rho$	-0.734	0.055	(-0.828, -0.615)
FIXEFS		mean	std.dev.	95% CRI
	INTERCEPT	-0.147	0.223	(-0.577, 0.292)
	<b>cond</b> -DE-ADV	1.548	0.219	(1.127, 1.979)
	<b>cond</b> -DE-OBJ	0.482	0.207	(0.078, 0.888)
	<b>cond</b> -DE-SUBJ	0.807	0.209	(0.396, 1.226)
	<b>cond</b> -NEGATIVE	3.152	0.268	(2.630, 3.689)
	<b>cond</b> -N-WORD-ADV	2.887	0.266	(2.379, 3.425)
	<b>cond</b> -N-WORD-OBJ	2.880	0.265	(2.359, 3.394)
	<b>cond</b> -N-WORD-SUBJ	2.775	0.253	(2.290, 3.281)
	<b>conx</b> -REV	-3.929	0.431	(-4.797, -3.125)
	<b>cond</b> -DE-ADV : <b>conx</b> -REV	1.882	0.407	(1.110, 2.694)
	<b>cond</b> -DE-OBJ : <b>conx</b> -REV	1.171	0.413	(0.376, 2.010)
	<b>cond</b> -DE-SUBJ : <b>conx</b> -REV	2.369	0.402	(1.595, 3.167)
	<b>cond</b> -NEGATIVE : <b>conx</b> -REV	2.386	0.457	(1.507, 3.312)
	<b>cond</b> -N-WORD-ADV : <b>conx</b> -REV	2.878	0.458	(2.002, 3.804)
	<b>cond</b> -N-WORD-OBJ : <b>conx</b> -REV	2.141	0.452	(1.291, 3.035)
	<b>cond</b> -N-WORD-SUBJ : <b>conx</b> -REV	3.244	0.454	(2.380, 4.157)

MCMC: 3 chains, 30000 iterations per chain, 5000 burn-in, 10 thinning.