Direct evidence of memory retrieval as a source of difficulty in non-local dependencies in language

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Direct evidence of memory retrieval as a source of difficulty in non-local dependencies in language

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Linguistic dependencies between non-adjacent words have been shown to cause comprehension difficulty, compared to local dependencies. According to one class of sentence comprehension accounts, non-local dependencies are difficult because they require the retrieval of the first dependent from memory when the second dependent is encountered. According to these memory-based accounts, making the first dependent accessible at the time when the second dependent is encountered should help alleviate the difficulty associated with the processing of non-local dependencies. In a dual-task paradigm, participants read sentences that did vs. didn’t contain a non-local dependency (i.e., object- and subject-extracted cleft constructions) while simultaneously remembering a word. The memory task was aimed at making the word held in memory accessible throughout the sentence. In an object-extracted cleft (e.g., *It was Ellen who John consulted…*), the object (*Ellen*) must be retrieved from memory when *consulted* is encountered. In the critical manipulation, the memory word was identical to the verb’s object (*ELLEN*). In these conditions, the extraction effect was reduced in the comprehension accuracy data and eliminated in the reading time data. These results add to the body of evidence supporting memory-based accounts of syntactic complexity.
Introduction

Interpreting syntactic dependencies among words is a critical component of language understanding. Although most such dependencies in human languages are between immediately adjacent elements (e.g., Hawkins, 1994; Collins, 1996; Ferrer i Cancho, 2006; Temperley, 2007; Gildea & Temperley, 2009; Park & Levy, 2009), in some cases words that need to be interpreted as dependent on one another are separated by other words or clauses. For example, in order to correctly interpret a sentence like (1), a dependency relationship needs to be established between girl and was, in spite of the presence of a plausible subject of was that is immediately adjacent to it (i.e., boy).

(1) The girl who kissed the boy was attractive.

Dependencies between words that are not adjacent to one another have been shown to cause processing difficulty, compared to cases where the dependencies are local, manifesting in slower online processing times and lower performance in the offline tasks that require reliance on the representation of the sentence’s content in several languages across many paradigms (e.g., English: Wanner & Maratsos, 1978; King & Just, 1991; Gibson, 1998, 2000; Gordon et al., 2001; McElree et al., 2003; Van Dyke & Lewis, 2003; Grodner & Gibson, 2005; Lewis et al., 2006; French: Holmes & O’Regan, 1981; Baudiffier, Caplan, Gaonac'h & Chesnet, 2011; German: Mecklinger, Schriefers, Steinhauer, & Friederici, 1995; Schriefers, Friederici, & Kuhn, 1995; Vasishth & Drenhaus, 2011, but see Konieczny, 2000; Konieczny & Doring, 2003; Levy & Keller, 2012; Dutch: Frazier, 1987; Mak, Vonk & Schriefers, 2002, 2006; Japanese: Miyamoto
& Nakamura 2003; Ishizuka, Nakatani & Gibson, 2003; Ueno & Garnsey, 2008; 
Korean: O’Grady, Lee & Choo, 2003; Kwon, Polinsky & Kluender, 2006; Kwon et al., 2010;
Russian: Levy, Fedorenko & Gibson, submitted; cf. Basque: Carreiras et al. 2010; Hindi: 
Vasishth & Lewis, 2006; Chinese: Hsiao & Gibson, 2003; Gibson & Wu, 2011, but see Lin & Bever 2006). The source of this difficulty is still under debate, however.

According to one class of accounts (memory-based accounts), non-local dependencies are difficult to process because they require the retrieval of the first dependent from memory when the second dependent is encountered. For example, in (1), girl has to be retrieved from memory when was is encountered.

Previous studies evaluating the predictions of memory-based accounts have typically relied on manipulations designed to make the retrieval of the non-local dependent more difficult, by increasing the distance between the two dependents, making the intervening words similar to the to-be-retrieved word, or including a secondary task designed to interfere with the retrieval of the target word (e.g., Gordon et al., 2001, 2002; Van Dyke & Lewis, 2003; Grodner & Gibson, 2005; Fedorenko, Gibson & Rohde, 2006; VanDyke & McElree, 2007). Greater processing difficulty with an increase in the difficulty of the retrieval operation has been interpreted as evidence for memory-based accounts. However, memory-based accounts make another prediction: making the retrieval of the non-local dependent easier at the point of dependency formation should reduce the difficulty associated with non-local dependencies. A couple of recent findings in the literature suggest that this might be the case. For example, Hofmeister (2011) demonstrated that increasing the representational richness of the to-be-retrieved word (e.g., by adding modifying adjectives or prepositional phrases) facilitates retrieval.
Relatedly, Vasishth et al. (in press) have observed that nouns in more prominent syntactic positions (clefted nouns) are easier to retrieve than their non-clefted counterparts in Hindi. The current experiment provides a more direct evaluation of this prediction of memory-based accounts.

**Experiment**

One well-studied contrast between local and non-local dependency structures involves subject- and object-extracted relative clauses or clefts, as in (2):

(2)  
   a. It was John who consulted Ellen in the library.  
   b. It was Ellen who John consulted in the library.

The object-extracted cleft (2b) contains a non-local dependency between the pronoun *who* (referring to *Ellen*) and the verb *consulted*. In contrast, in the subject-extracted cleft (2a), *consulted* is immediately adjacent to both its subject *who* (referring to *John*) and its object *Ellen*.

Sentences containing subject- vs. object-extractions are commonly used for investigating structural dependencies because they differ only in the positions of the critical words, so that the critical comparisons are performed on similar (often identical) words that are located in similar positions in the sentence. Furthermore, it is easy to match such materials for plausibility, either by fixing the meaning across the two structures as in (2), or by including two versions of each structure (e.g., using *It was Ellen who consulted John in the library* and *It was John who Ellen consulted in the library* in
addition to the versions in (2)). We therefore chose to use subject- and object-extracted cleft structures in the current experiment.

To achieve the desired manipulation – facilitating the retrieval of the non-local dependent at the point of dependency formation – we adapted a dual-task paradigm in which participants see a word or a set of words that they are instructed to remember, then read a sentence word-by-word, and then report the word(s) held in memory and answer a question about the sentence (e.g., Gordon et al., 2002; Fedorenko et al., 2006). We made two modifications to the earlier versions of this paradigm. First, we made the memory task easy by having participants keep at most one word in memory. And second, in the critical condition, we made the word held in memory identical to the object noun. If the difficulty in the object-extracted structures is due to the need to retrieve the object noun from memory, then making the object noun salient throughout the sentence (via the memory task) – thereby eliminating the need to retrieve it from memory at the verb – should partially or completely alleviate the difficulty typically associated with the processing of object-extracted structures.

**Methods**

**Participants** Sixty native English speakers between 18 and 40 years old – students at MIT and members of the surrounding community – participated in the study. All participants were paid for their participation and were naive to the purposes of the study.

**Design and materials** The experiment manipulated the cleft structure (subject-extracted, object-extracted) and the presence and kind of the word held in memory (identical to the object of the embedded verb, a control noun that did not appear in the sentence, identical
to the subject of the embedded verb, or no word to remember) in a 2x4 design, as schematically illustrated in (3). The critical prediction concerns the object-memory-word object-extracted condition, where the object noun (Ellen) should no longer be difficult to retrieve at the point of processing the verb because the object is made salient throughout the sentence and should therefore be more easily accessible when the verb-object dependency is formed. This manipulation should therefore lead to faster processing times and a more robust memory representation of the sentence, compared to the no-memory-word condition.

The control-memory-word conditions were included in order to determine the effect on the sentence-reading task of the secondary (memory) task, thus providing an additional baseline for the critical object-memory-word conditions. This manipulation should make the object-extracted condition (and possibly the subject-extracted condition) more difficult than the no-memory-word conditions (Gordon et al., 2002; Fedorenko et al., 2006).

Finally, the subject-memory-word conditions were included to evaluate an alternative hypothesis about why processing a sentence like It was Ellen who John consulted in the library might be facilitated when the word ELLEN is held in memory. In particular, Ellen is the topic of the sentence, and having the memory word match the topic may facilitate the topic’s encoding leading to a stronger memory trace and a facilitatory effect on the processing of the rest of the sentence. According to this hypothesis, in addition to the facilitation in the object-memory-word object-extracted condition, we should see faster processing times and higher comprehension accuracies in
the subject-memory-word subject-extracted condition compared to the subject-extracted condition where no word is held in memory.

Note that both a) the critical hypothesis about the facilitated retrieval of the object noun from memory in the object-memory-word object-extracted condition, and b) the alternative hypothesis about the general facilitation resulting from a more robust encoding of the sentence topic (which predicts facilitation in both the object-memory-word object-extracted condition and in the subject-memory-word subject-extracted condition), appeal to memory processes. However, whereas the latter focuses on the encoding of the clefted noun, the former focuses on its retrieval at the point of dependency formation (see e.g., Lewis et al., 2006, and Hofmeister, 2011, for a discussion of encoding/retrieval processes in sentence comprehension). It is also worth noting that the two hypotheses are not mutually exclusive.

(3)

<table>
<thead>
<tr>
<th>Condition</th>
<th>Memory word</th>
<th>Sentence</th>
</tr>
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<tbody>
<tr>
<td>No memory word / Subj-extracted</td>
<td>X</td>
<td>It was John who consulted Ellen in the library.</td>
</tr>
<tr>
<td>No memory word / Obj-extracted</td>
<td>X</td>
<td>It was Ellen who John consulted in the library.</td>
</tr>
<tr>
<td>Obj. mem. word / Subj-extracted</td>
<td>ELLEN</td>
<td>It was John who consulted Ellen in the library.</td>
</tr>
<tr>
<td>Obj. mem. word / Obj-extracted</td>
<td>ELLEN</td>
<td>It was Ellen who John consulted in the library.</td>
</tr>
<tr>
<td>Ctrl mem. word / Subj-extracted</td>
<td>STEVE</td>
<td>It was John who consulted Ellen in the library.</td>
</tr>
<tr>
<td>Ctrl mem. word / Obj-extracted</td>
<td>STEVE</td>
<td>It was Ellen who John consulted in the library.</td>
</tr>
<tr>
<td>Subj. mem. word / Subj-extracted</td>
<td>JOHN</td>
<td>It was John who consulted Ellen in the library.</td>
</tr>
<tr>
<td>Subj. mem. word / Obj-extracted</td>
<td>JOHN</td>
<td>It was Ellen who John consulted in the library.</td>
</tr>
</tbody>
</table>
Forty sets of sentences were constructed (see Appendix). Each participant saw only one version of each item, following a Latin-Square design. Names were used instead of occupation nouns – often used in linguistic materials – for three reasons. First, using names largely eliminates the issues with potential differences in plausibility between the subject- and object-extracted conditions. In particular, in the sentences where a name is used in both the subject and the object position the only cue to the dependency structure is the syntax of the sentence (i.e., the positions of the nouns relative to the verb and to each other in this case). Second, somewhat relatedly, if we used occupation nouns, we would have to control for the potential semantic relationships between the nouns (both sentence and memory nouns) and the verb. The effect of the memory word on the processing of the sentence may therefore be due not only to the relationship between the memory word and the sentence nouns, but also to the relationship between the memory word and the sentence verb. Using names allowed us to examine retrieval difficulty as a function of how accessible in memory the target noun is, without worrying about the verb’s semantics. Finally, using names reduces the potential variability that may be associated with the familiarity/frequency of different common nouns (such as occupations), allowing us to focus on the manipulation of interest, i.e., on the dependency structure of the sentences. Because we used names, we had to use the cleft construction instead of the relative clause construction because only the former allows the use of names in the head noun position.

In constructing the materials, we had to make a choice between (a) keeping the words in the critical region (the embedded noun and the verb) identical across the
subject- and the object-extracted conditions, and (b) keeping the memory word identical – within each memory-word condition – across the subject and the object-extracted conditions. We chose the latter because we thought it was more important to compare the effects of the same memory word on the processing of sentences that have the same meaning but differ in structure. To control for potential issues of comparing across different words in the subject- and object-extracted conditions, each name appeared once in each syntactic position across the item set. For example, in addition to the item in (2) where John is used in the subject position, there was another item where John is used in the object position. Similarly, there was another item where Ellen is used in the subject position.

Verbs were chosen such that a male or a female was similarly likely to be the agent or the patient. The gender of the names was counterbalanced (ten items used two male names; ten – two female names; ten – male subject and female object; and ten – female subject and male object). For the sentences where both nouns were either male or female, the control memory word matched the sentence nouns in gender. For the sentences where one name was male and one female, the control memory word was male half of the time.

To assess the robustness of the memory representations of the sentence content, a yes/no comprehension question about the propositional content of the sentence was asked at the end of each trial. Two question types were used, each requiring the understanding of the dependency structure of the sentence: (a) a question asking only about the dependency structure (e.g., Did John consult Ellen? Yes / Did Ellen consult John? No, for the item in (2)); and (b) a question asking about both the dependency structure and the
sentence ending (e.g., Did John consult Ellen in the library? Yes / Did Ellen consult John in the classroom? No). Within an item, the same question was asked for all the conditions. Question types varied across items and the number of “yes” and “no” questions was balanced.

The experiment also included 80 filler sentences containing names, which were similar to the critical sentences in length and complexity. In half of the fillers, the memory word was a name that appeared in the sentence, in a quarter of the fillers, the memory word was a name that did not appear in the sentence, and in the remaining quarter there was no memory word, to match the distribution of memory words in the critical sentences. (The fillers are available from the authors upon request.)

Trials within each experimental list were randomized separately for each participant.

**Procedure** The sentence-reading task used self-paced moving-window word-by-word reading (Just, Carpenter & Woolley, 1982). The experiment was run using the Linger 2.85 software by Doug Rohde.

Each trial began with a word or an “X” appearing on the screen in capital letters for 600 msec. An “X” was used for the no-memory-word trials, to match the structure of the trials in the other conditions. Participants were instructed to remember the word. A blank screen was then presented for 500 msec, which was followed by a series of dashes marking the length and position of the words in the sentence. Participants pressed the spacebar to reveal each word of the sentence. As each new word appeared, the preceding word was converted to dashes again. The amount of time participants spent reading each word was recorded as the time between key presses. After the sentence, a box appeared,
and participants were instructed to type in the word from the memory task or an “X” (for the no-memory-word trials). Then a comprehension question appeared on the screen. Participants pressed one of two keys to respond “yes”/“no”. At the end of each trial, participants were shown the percentage of correct responses in the experiment so far, across the memory and the comprehension tasks. They were told to take low percentages as an indication to be more careful.

Before the experiment, a short list of practice items was presented to familiarize participants with the task. The experiment took approximately 40 minutes.

Results
Before conducting detailed analyses, we examined individual subjects’ accuracies on the memory task and the comprehension question task. We removed four subjects with accuracies lower than 67% on either the memory task or the comprehension question task. For the remaining 56 subjects, the across-conditions mean was 96.7% for the memory task, and 90.7% for the comprehension question task.

Analyses reported here were conducted with the lme4 package (Bates et al., 2008) for the statistical language R (R Core Development Team, 2008). The data from the memory task and the comprehension task were analyzed with logistic regressions. The reading time data were analyzed with a linear mixed-effects regression. Recent results have shown that including only random intercepts in linear mixed-effects regressions can be anti-conservative (e.g., Barr et al., submitted), so we also include random slopes for participants in modeling the reading time data. (Further inclusion of random slopes for items led to a lack of model convergence.) Significance (p) values were estimated from (a) the t-values that were obtained from the lmer function; and (b) conservative estimates
of the number of degrees of freedom in the model. The estimates of the number of
degrees of freedom in the model consisted of the number of observations minus the
number of intercepts fit in the model (the number of participants + the number of items =
56+40 = 96) and the number of slopes being fit in the model (the number of participants
= 56).

**Memory task performance** As intended, the memory task was easy, as evidenced by
close-to-ceiling performance\(^1\) (Table 1). We performed a logistic regression with two
sum-coded factors: extraction (subject-, object-) and type of memory word (object
memory word, control memory word, subject memory word). The control-memory-word
object-extracted condition was chosen as the baseline. The results of the model are
summarized in Table 2. The extraction manipulation did not have a significant effect on
accuracy. For the memory word manipulation, the subject-memory-word conditions
were not reliably different from the baseline, but the object-memory-word conditions
were reliably better. No interactions were observed. The small but reliable improvement
in the object-memory-word conditions, compared to the control-memory-word
conditions, is plausibly due to the overall better performance on the object-memory-word
conditions, as discussed in the sections below.

**Comprehension accuracies** Accuracies for the eight conditions are shown in Figure 1.
We performed a logistic regression with two sum-coded factors: extraction (subject-,
object-) and type of memory word (no memory word, object memory word, control
memory word, subject memory word). The no-memory-word object-extracted condition
was chosen as the baseline. The results of the model are summarized in Table 3. The

\(^1\) Deviations from the target by one letter were not counted as errors.
extraction manipulation had a highly reliable effect on accuracy. For the memory word manipulation, we observed two main effects: the object-memory-word conditions were reliably more accurate, and the control-memory-word conditions were reliably less accurate than the baseline (no-memory-word conditions). Lower accuracies in the control-memory-word conditions are consistent with previous findings (e.g., Gordon et al., 2002; Fedorenko et al., 2006). Furthermore, we observed a significant interaction, such that the extraction effect was smaller for the object-memory-word conditions compared to the no-memory-word conditions.

**Reading times** Before performing linear mixed-effects regression analyses on the reading time data, we removed raw reading times that were longer than 5,000 msec and then transformed the raw reading times into residual reading times (Ferreira & Clifton, 1986; see Trueswell, Tanenhaus, & Garnsey, 1994, for discussion). First, a regression equation predicting reading times from region length was derived for each participant, using all filler and target items. Then, for each region, the reading time predicted by the participant’s regression equation was subtracted from the actual measured reading time to obtain a residual reading time. We analyzed residual reading times because (a) they adjust for differences in region lengths and overall differences in participants’ reading rates; and (b) they are more normally distributed than raw reading times (normally distributed data is a requirement for regression analyses, cf. Jaeger, 2008). Residual reading times greater than three standard deviations away from the mean for a position within condition were removed from the analyses. These procedures resulted in the removal of 3.2% of the data.
We defined the critical region as the portion of the sentence where the dependency structure is manipulated (*consulted Ellen / John consulted* in (2)), as is commonly done in the literature. Although the critical effects are predicted to occur at the verb (*consulted*), this word alone is not directly comparable across the subject- and the object-extracted conditions because (a) the verb occurs in different sentence positions and position within the sentence is known to affect reading times, and, even more importantly, (b) in the subject-extracted conditions at the point where the verb is processed the object noun has not yet been encountered and therefore the memory word manipulation in the subject-memory-word and control-memory-word conditions has not yet had a chance to affect sentence processing. Figure 2 presents the RTs for the critical region across the eight conditions (see Appendix for the sentence word-by-word RT graphs).

We performed a linear mixed-effects regression with two sum-coded factors: extraction (subject-, object-) and type of memory word (no memory word, object memory word, control memory word, subject memory word). As in the analysis of comprehension accuracies, the no-memory-word object-extracted condition was chosen as the baseline. In order to remove the potential contribution of the facilitated encoding of the clefted noun, reading times on the region preceding the critical region (i.e., *who*) were included as an additional predictor in the model, per a reviewer’s suggestion. The results of the model are summarized in Table 4. The extraction manipulation (as well as the reading times on the preceding region) had highly reliable effects on reading times (both \(ps<.005\)). No main effects were observed for the memory word manipulation. However, a significant interaction between extraction and memory word manipulations
was observed, such that the extraction effect was smaller for the object-memory-word conditions compared to the no-memory-word conditions ($p<.05$). There was additionally a marginal interaction such that the extraction effect was larger for the subject-memory-word conditions compared to the no-memory-word conditions ($p=.084$).

**Discussion**

We here tested a prediction of memory-based accounts of syntactic complexity. According to these accounts, non-local dependencies are more costly than local dependencies because in the former the first element of the dependency must be retrieved from memory when the second element of the dependency is encountered (unlike local dependencies, where the first element of the dependency is still highly active when the second element is encountered). These accounts therefore predict that making the to-be-retrieved element highly active in memory at the point when the dependency is formed should reduce the difficulty typically associated with non-local dependencies.

We used a dual-task paradigm in an attempt to facilitate the retrieval of the object noun from memory in object-extracted cleft constructions by making the object noun highly accessible at the point of the dependency formation. For example, participants were holding in memory the word *ELLEN* while processing the sentence *It was Ellen who John consulted at the library*. When *consulted* is encountered and the dependency between *consulted* and its object (*Ellen*) needs to be established, the object noun is already highly active in memory. Consistent with the prediction of memory-based accounts, holding in memory the object noun during the processing of object-extracted conditions led to a robust memory representation of the dependency structure:
comprehension question accuracies were higher in this condition than in the other three object-extracted conditions (even in the no-memory-word condition, which had no memory-task demands). A similar pattern was observed in the reading times at the critical region: reading times in the object-memory-word condition were numerically faster than in the other three object-extracted conditions, and the extraction effect was eliminated in the object-memory-word conditions.

What is the nature of the facilitation effect observed in the object-memory-word object-extracted condition? As discussed in Methods, an alternative to the retrieval facilitation hypothesis is a hypothesis whereby a match between the memory word and the topic of the sentence facilitates the encoding of the topic noun, leading to a facilitatory effect on the processing of the sentence. This hypothesis predicts faster processing times and higher comprehension accuracies not only in the object-memory-word object-extracted condition but also in the subject-memory-word subject-extracted condition.

The reading time data patterns provide some support for the encoding-based hypothesis. In particular, at the clefted noun region, the subject-memory-word condition is numerically the fastest of the four subject-extracted conditions, and the object-memory-word condition is the fastest of the four object-extracted conditions (see Appendix). Furthermore, at the critical region, the conditions where the memory word mismatches the topic are slower than those where the memory word matches the topic (or where no word is held in memory). Importantly, however, in our analysis of the reading time data we included reading times on the preceding region (who) as a predictor in the model, and the critical effects remained significant. Moreover, the comprehension
accuracy data pattern is consistent with the retrieval-based, but not the encoding-based, hypothesis. In particular, the subject-memory-word subject-extracted condition is less, not more, accurate than the no-memory-word subject-extracted condition, and the subject-memory-word object-extracted condition is more, not less, accurate than the no-memory word object-extracted condition. As a result, we tentatively conclude that the effect observed in the object-memory-word object-extracted condition is largely due to facilitated retrieval of the object noun, but leave open the possibility of enhanced encoding playing some role, too. As noted above, these hypotheses are not mutually exclusive.

Until now we have been focusing on memory-based accounts of syntactic complexity. According to an alternative class of accounts, the difficulty in non-local dependencies has to do with the lower frequency of these structures in the input, compared to local-dependency structures (e.g., Keenan & Comrie, 1977; Hale, 2001; Reali & Christiansen, 2007; Gennari & MacDonald, 2008; Levy, 2008). Whereas much existing evidence concerning subject- and object-extracted structures is consistent with both memory-based and experience-based accounts, some evidence is better explained by one or the other. Experience-based accounts are better able to explain the effects of position-dependent lexico-semantic properties of the noun phrases on processing difficulty. For example, the difficulty of processing object-extractions is affected by whether the subject or the object is animate vs. inanimate (e.g., Traxler, Morris & Seeley, 2002; but see Gibson et al., in press; Tily et al., submitted), or whether the subject or the object is a pronoun (e.g., Warren & Gibson, 2002; Reali & Christiansen, 2007): the less frequent configurations lead to more processing difficulty.
On the other hand, memory-based accounts are better able to explain (1) the locus of the processing difficulty in object-extracted structures, and (2) similarity-based interference effects in normal reading and in dual-task paradigms. First, most of the difficulty in object-extracted structures occurs at the verb (cf. Staub, 2010). This is straightforwardly explained by memory-based accounts because the verb is where the dependency is formed and where therefore the retrieval of the object noun from memory takes place. In contrast, experience-based accounts predict processing difficulty to occur as soon as the comprehender can determine that an object-extracted structure has been encountered, which is at the embedded subject. However, little, if any, processing difficulty is experienced during the embedded subject or during the material that can be inserted between the embedded subject and embedded verb (Grodner & Gibson, 2005; Gibson, Tily & Fedorenko, in press, cf. Staub, 2010). And second, the difficulty of processing object-extractions increases when the subject and the object are syntactically and semantically similar (Gordon et al., 2001, 2004), or when participants have to hold in memory words that are similar to the nouns in the sentence (Gordon et al., 2002; Fedorenko et al., 2006) or that compete with the object noun for being interpreted as the object of the verb (Van Dyke & McElree, 2007). These results are predicted by memory-based accounts, but not by experience-based accounts. The results reported here provide new direct evidence in support of memory-based accounts, which cannot be straightforwardly explained by experience-based accounts. As many researchers now concur, features of both memory-based and experience-based accounts are necessary to explain the wealth of the available evidence (e.g., Lewis et al., 2006; Demberg & Keller, 2008; Boston et al., 2011; Gibson et al., in press).
## Tables and figures

### Memory word

<table>
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<tr>
<th>Structure</th>
<th>Object</th>
<th>Subject</th>
<th>Control</th>
</tr>
</thead>
<tbody>
<tr>
<td>Subject-extracted</td>
<td>.986 (.01)</td>
<td>.982 (.01)</td>
<td>.961 (.01)</td>
</tr>
<tr>
<td>Object-extracted</td>
<td>.989 (.01)</td>
<td>.968 (.01)</td>
<td>.918 (.02)</td>
</tr>
</tbody>
</table>

Table 1. Memory task performance in the six conditions (standard errors of the by-participants mean in the parentheses).

Coefficients:

|                      | Estimate | Std. Error | z value | Pr(>|z|) |
|----------------------|----------|------------|---------|---------|
| (Intercept)          | 3.62909  | 0.17087    | 21.239  | < 2e-16 *** |
| Extraction.subj      | 0.18242  | 0.17087    | 1.068   | 0.28568 |
| MemWord.object       | 0.75067  | 0.28002    | 2.681   | 0.00734 ** |
| MemWord.subject      | 0.07703  | 0.23608    | 0.326   | 0.74421 |
| Extraction.subj:MemWord.object | -0.32807 | 0.28002    | -1.172  | 0.24135 |
| Extraction.subj:MemWord.subject | 0.11879  | 0.23608    | 0.503   | 0.61482 |

Significance codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Table 2. Results of the logistic regression on the memory task data.

Coefficients:

|                      | Estimate | Std. Error | z value | Pr(>|z|) |
|----------------------|----------|------------|---------|---------|
| (Intercept)          | 2.473440 | 0.086913   | 28.459  | < 2e-16 *** |
| Extraction.subj      | 0.613683 | 0.086913   | 7.061   | 1.65e-12 *** |
| MemWord.subject      | 0.020672 | 0.149431   | 0.138   | 0.88997 |
| MemWord.object       | 0.335753 | 0.157708   | 2.129   | 0.03326 * |
Table 3. Results of the logistic regression on the comprehension question data.

|                           | Estimate | Std. Error | t value | Pr(>|t|) |
|---------------------------|----------|------------|---------|---------|
| MemWord.control           | -0.437014| 0.134554   | -3.248  | 0.00116 ** |
| Extraction.subj MemWord.subject | -0.001714| 0.149431   | -0.011  | 0.99085   |
| Extraction.subj MemWord.object | -0.316795| 0.157708   | -2.009  | 0.04456 *  |
| Extraction.subj MemWord.control | 0.085023 | 0.134554   | 0.632   | 0.52746   |

Significance codes:   0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Table 4. Results of the logistic regression on the comprehension question data.

Coefficients:

|                           | Estimate | Std. Error | t value | Pr(>|t|) |
|---------------------------|----------|------------|---------|---------|
| (Intercept)               | -29.71625| 19.24355   | -1.544  | 0.124    |
| PrecRegion.RT            | 0.69848  | 0.05971    | 11.697  | 0.000    ** |
| Extraction.subj          | -29.36982| 9.09843    | -3.228  | 0.001 **  |
| Related.sumCoding1       | 7.63745  | 10.88957   | 0.701   | 0.484    |
| Related.sumCoding2       | -2.02203 | 10.93088   | -0.185  | 0.854    |
| Related.sumCoding3       | 12.48072 | 12.34064   | 1.011   | 0.313    |
| Extraction.subj MemWord.subject | -25.52178| 14.43754   | -1.768  | 0.078    |
| Extraction.subj MemWord.object | 22.00434 | 10.21489   | 2.154   | 0.035 *  |
| Extraction.subj MemWord.control | -3.47168 | 11.92366   | -0.291  | 0.769    |
Figure 1. Comprehension question accuracies in the eight conditions. Error bars represent standard errors of the by-participants mean.

Figure 2. Reading times at the critical region in the eight conditions. Error bars represent standard errors of the by-participants mean.
Appendix

Experimental materials

We first demonstrate how the eight conditions were created using the first item. For the rest of the items we show the subject-extracted version and provide the control memory noun. The rest of the conditions can be generated as shown for item #1.

1. It was John who consulted Ellen in the library. [STEVE]

Subject-extracted conditions:
   No memory word: [X] It was John who consulted Ellen in the library.
   Object memory word: [ELLEN] It was John who consulted Ellen in the library.
   Control memory word: [STEVE] It was John who consulted Ellen in the library.
   Subject memory word: [JOHN] It was John who consulted Ellen in the library.

Object-extracted conditions:
   No memory word: [X] It was Ellen who John consulted in the library.
   Object memory word: [ELLEN] It was Ellen who John consulted in the library.
   Control memory word: [STEVE] It was Ellen who John consulted in the library.
   Subject memory word: [JOHN] It was Ellen who John consulted in the library.

2. It was Sara who distracted George at the meeting. [EMILY]
3. It was Ellen who offended Sara during the class. [JENNIFER]
4. It was George who criticized John after the talk. [MATT]
5. It was Edward who helped Caroline in the park. [ANNA]
6. It was Jennifer who congratulated Steve at the show. [JEFFREY]
7. It was Caroline who interviewed Jennifer in the office. [SOPHIE]
8. It was Steven who called Edward from the station. [JACOB]
9. It was Dan who emailed Sally about the job. [MICHAEL]
10. It was Jill who noticed Pete at the supermarket. [MEG]
11. It was Sally who applauded Jill after the debate. [RACHEL]
12. It was Pete who disliked Dan after the incident. [DAVE]
13. It was Joe who hired Laura for the position. [REBECCA]
14. It was Melissa who rewarded Jim with the bonus. [GREG]
15. It was Laura who tutored Melissa before the exam. [ABIGAIL]
16. It was Jim who followed Joe to the building. [ETHAN]
17. It was Noah who kicked Emma in the shin. [ALEX]
18. It was Olivia who neglected Aaron at the party. [HANNAH]
19. It was Emma who harassed Olivia in the bookstore. [LILY]
20. It was Aaron who misinformed Noah about the deal. [ANDREW]
21. It was Jack who misquoted Grace in the article. [ISABELLE]
22. It was Elizabeth who admired Nick for his courage. [JOSH]
23. It was Grace who warned Elizabeth about the storm. [AMELIA]
24. It was Nick who chased Jack around the lake. [OWEN]
25. It was Zachary who blamed Alyssa for the mistake. [WILLIAM]
26. It was Allison who thanked Adam for the gifts. [ABBY]
27. It was Alyssa who picked Allison for the team. [CLAIRE]
28. It was Adam who chaperoned Zachary to the dance. [IAN]
29. It was Colin who greeted Sophie at the door. [NATALIE]
30. It was Eva who punished James for his actions. [RYAN]
31. It was Sophie who attacked Eva at the conference. [SAMANTHA]
32. It was James who confused Collin during the discussion. [SETH]
33. It was Nathan who defended Lauren at the interview. [SAMUEL]
34. It was Amy who accused Tony at the hearing. [AMBER]
35. It was Lauren who provoked Amy during the argument. [LEAH]
36. It was Tony who stabbed Nathan on the street. [EVAN]
37. It was Justin who visited Erin at her house. [LUCY]
38. It was Victoria who inspired Tyler at the gallery. [BRENDAN]
39. It was Erin who surprised Victoria at the airport. [JULIA]
40. It was Tyler who described Justin to the police. [HENRY]

Residual reading times across the sentence

![Graph showing residual reading times across the sentence](image-url)
Acknowledgements

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