

Dependency resolution difficulty increases with distance in Persian separable complex predicates: Implications for expectation and memory-based accounts

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Keywords

locality, expectation, Persian, Complex predicate, self-paced reading, surprisal, entropy

Abstract

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Delaying the appearance of a verb in a noun-verb dependency tends to increase processing difficulty at the verb; one explanation for this locality effect is decay and/or interference of the noun in working memory. Surprisal, an expectation-based account, predicts that delaying the appearance of a verb either renders it no more predictable or more predictable, leading respectively to a prediction of no effect of distance or a facilitation. Recently, Husain et al (2014) suggested that when the exact identity of the upcoming verb is predictable (strong predictability), increasing argument-verb distance leads to facilitation effects (consistent with surprisal), but when the exact identity of the upcoming verb is not predictable (weak predictability), locality effects are seen. We investigated Husain et al's proposal using Persian complex predicates (CPs), which consist of a non-verbal element ('noun' in the current study) and a verb. In such constructions, once the noun has been read, the exact identity of the verb is highly predictable (strong predictability); this was confirmed using a sentence completion study. In two self-paced reading (SPR) and two eye-tracking (ET) experiments, we delayed the appearance of the verb by interposing a relative clause (Expt. 1 and 3) or a long PP (Expt. 2 and 4).

We also included a simple predicate (Noun-Verb) configuration with the same distance manipulation; here, the exact identity of the verb was not predictable (weak predictability). Thus, the design crossed Predictability Strength and Distance. We found that, consistent with surprisal, the verb in the strong predictability conditions was read faster than in the weak predictability conditions. Furthermore, greater verb-argument distance led to slower reading times; strong predictability did not neutralize or attenuate the locality effects. As regards the effect of distance on dependency resolution difficulty, these four experiments present evidence in favor of working memory accounts of argument-verb dependency resolution, and against the surprisal-based expectation account of Levy (2008). However, another expectation-based measure, entropy (which was computed using the sentence completion data) predicts reading times in Experiment 1 (but not the other experiments). Thus, memory overload and entropy are two alternative explanations for the locality effects in Persian.

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In Germany, the Principal Investigator (Shravan Vasishth) is responsible for ensuring that ethical standards are adhered to. All studies at Potsdam are conducted in accordance with the Helsinki declaration.

Please detail the consent procedure used for human participants or for animal owners. If not applicable, please state this.
Not applicable.

Please detail any additional considerations of the study in cases where vulnerable populations were involved, for example minors, persons with disabilities or endangered animal species. If not applicable, please state this.

Not applicable.

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2 ABSTRACT

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5 the noun in working memory. Surprisal, an expectation-based account, predicts that delaying
6 the appearance of a verb either renders it no more predictable or more predictable, leading
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8 suggested that when the exact identity of the upcoming verb is predictable (strong predictability),
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11 effects are seen. We investigated Husain et al's proposal using Persian complex predicates
12 (CPs), which consist of a non-verbal element ('noun' in the current study) and a verb. In such
13 constructions, once the noun has been read, the exact identity of the verb is highly predictable
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19 crossed Predictability Strength and Distance. We found that, consistent with surprisal, the
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21 conditions. Furthermore, greater verb-argument distance led to slower reading times; strong
22 predictability did not neutralize or attenuate the locality effects. As regards the effect of distance
23 on dependency resolution difficulty, these four experiments present evidence in favor of working
24 memory accounts of argument-verb dependency resolution, and against the surprisal-based
25 expectation account of Levy (2008). However, another expectation-based measure, entropy
26 (which was computed using the sentence completion data) predicts reading times in Experiment
27 1 (but not the other experiments). Thus, memory overload and entropy are two alternative
28 explanations for the locality effects in Persian.

29 **Keywords:** Locality, Expectation, Surprisal, Entropy, Persian, Complex Predicates, self-paced reading, eye-tracking

1 INTRODUCTION

30 A long-standing claim in sentence processing is that increasing distance in a linguistic dependency, such
 31 as a noun-verb dependency, leads to greater processing difficulty (**Chomsky**, 1965; **Just and Carpenter**,
 32 1992; **Gibson**, 2000; **Lewis and Vasishth**, 2005); it is common to refer to this increase in processing
 33 difficulty as the locality effect. One explanation for the locality effect is in terms of constraints imposed
 34 by working memory. According to one account, the Dependency Locality Theory (DLT; **Gibson** (1998)),
 35 the processing difficulty experienced when resolving a long dependency depends on the decay experienced
 36 by the noun; a related account by **Lewis and Vasishth** (2005) attributes the locality effect to decay and/or
 37 interference. Constraints on working memory may be a plausible explanation given that individuals'
 38 working memory capacity seems to affect the processes involved in dependency resolution (**Nicenboim**
 39 **et al.**, 2015; **Caplan and Waters**, 2013). Although there is evidence consistent with the memory-based
 40 explanation in English, German, Chinese, Russian, and Hindi, (**Hsiao and Gibson**, 2003; **Grodner and**
 41 **Gibson**, 2005; **Bartek et al.**, 2011; **Vasishth and Drenhaus**, 2011; **Levy et al.**, 2013; **Husain et al.**,
 42 2014, 2015), research on some of these languages has also uncovered evidence that increasing noun-verb
 43 distance facilitates processing at the verb (**Konieczny**, 2000; **Vasishth**, 2003; **Vasishth and Lewis**, 2006;
 44 **Jaeger et al.**, 2008; **Vasishth and Drenhaus**, 2011; **Levy and Keller**, 2013; **Husain et al.**, 2014; **Jäger**
 45 **et al.**, 2015). One explanation for these anti-locality effects is in terms of surprisal (**Hale**, 2001; **Levy**,
 46 2008). Surprisal extends and formalizes the old idea of predictive sentence processing—which has been
 47 extensively investigated in the EEG literature (e.g., **Kutas and Hillyard** 1984)—in terms of probabilistic
 48 parse continuations (also see **Jurafsky** 1996). The surprisal account assumes that the comprehender
 49 maintains and uses linguistic knowledge probabilistically to parse a sentence incrementally. Surprisal
 50 is the claim that rare transitions are difficult: increased processing difficulty is predicted when a parser is
 51 required to build a low-probability syntactic structure. Formally, surprisal is defined as the negative log
 52 probability of encountering a particular part of speech or word given previous context. We will refer to
 53 surprisal as the expectation-based account, following the terminology of **Levy** (2008).¹

54 In many of these studies, evidence has been found for both the memory-based account and the
 55 expectation-based account. One conclusion that has emerged is that both memory and expectation play
 56 a role. For example, in his eye-tracking study investigating processing difference in English object vs
 57 subject relative clauses, **Staub** (2010) finds evidence for both expectation-based processing and locality
 58 constraints, although these occur in different regions of the target sentence. An example of Staub's design
 59 is provided below. In this study, processing difficulty was found on the noun phrase *the fireman* in the
 60 ORC (object relative clause) 1b, compared to the SRC (subject relative clause) 1a; this is consistent with
 61 the expectation account because the reader would be forced to build a rare object relative in the ORC
 62 condition when he/she encounters the noun phrase. However, this study also found greater processing
 63 difficulty at the relative clause verb in ORCs than SRCs, which is predicted by memory accounts.

- 64 (1) a. The employees that noticed the fireman hurried across the open field.
 65
 66 b. The employees that the fireman noticed hurried across the open field.
 67

68 As further examples, both **Vasishth and Drenhaus** (2011) and **Levy and Keller** (2013) have argued
 69 that locality effects may appear when high working memory load is experienced; anti-locality effects may
 70 be present when the load is low.

¹ Another expectation-based account in the literature is the entropy reduction hypothesis or ERH (**Hale**, 2006); we do not investigate ERH in this paper, but we will briefly discuss a related idea, entropy, in the General Discussion.

71 In a recent development, **Husain et al.** (2014) argue that the strong predictability for a head
 72 (predicting an exact lexical item) can neutralize the locality effect; locality may manifest itself only
 73 when predictability strength is weak, that is, when only a verb phrase is predicted, and not the exact
 74 identity of the verb. In their self-paced reading study, **Husain et al.** (2014) used a 2×2 design, crossing
 75 Predictability and Dependency Distance to investigate locality and anti-locality effects. In the strong
 76 predictability conditions, Hindi complex predicates were used. In these noun-verb sequences, the noun
 77 strongly predicted the upcoming light verb, e.g. the noun *khayaal*, ‘care’, strongly predicts the verb
 78 *rakhnaa*, ‘put’, in *khayaal rakhnaa*, literally, ‘care put’ (‘to take care of’). The weak predictability
 79 condition, on the other hand, used the same verb used in the complex predicate, but the noun did not
 80 predict the verb. An example is *gitaar rakhnaa*, ‘guitar put’; ‘to put (down) a guitar’; here, the verb retains
 81 its literal meaning. Thus, when the reader see *gitaar*, they cannot predict the exact identity of the verb,
 82 because many other verbs are possible here (e.g., bought). To summarize, in the strong predictability
 83 condition, the noun predicted the exact identity of the verb, while in the weak predictability condition
 84 the exact identity of the verb was not predicted with high certainty—although a verb was predicted.
 85 The second factor, dependency distance, was manipulated by placing one to two adverbials between
 86 the nominal predicate/object and the verb in the short condition. The long condition had two to three
 87 intervening adverbials. Reading time was measured at the verb. The results showed that CP light verbs
 88 were read faster in long vs short distance conditions, but for non-CP verb there was a tendency towards
 89 a slowdown in long vs short conditions. Finally, there was weak evidence for an interaction (estimate on
 90 the log ms scale: 0.03, Bayesian 95% credible interval [-0.02, 0.07], posterior probability of the effect
 91 being greater than 0 was 0.77). That is, there was some evidence that with increased distance there was
 92 a speedup at the light verb in the CP conditions and a slowdown in the non-CP conditions. These results
 93 were interpreted by Husain and colleagues as strong predictability of the head canceling the locality effect,
 94 and the locality effect manifesting itself only when predictability strength was weak.

95 In the present study, we build on the work by **Husain et al.** (2014) described above. Husain and
 96 colleagues’ work suggested that the strength of the predictability may modulate whether locality effects
 97 occur or not; we investigate the cross-linguistic generality of this claim using Persian, which, like Hindi,
 98 also has a complex predicate construction that allows us to manipulate strong and weak predictability.
 99 We turn next to a short discussion of the complex predicate construction in Persian as it relates to our
 100 experiments.

2 COMPLEX PREDICATES IN PERSIAN

101 Complex Predicates (CPs) (**Samvelian**, 2001) consist of a sequence containing a non-verbal element
 102 (often a noun) and a verb, where the meaning of the sequence is non-compositional. An example is shown
 103 in (2).

- 104 (2) Maryam be man latme zad
 Maryam to me damage hit
 105 ‘Maryam caused damage to me (Maryam harmed me).’

106 The verb, often called a light verb, lacks sufficient semantic force to function as an independent predicate
 107 (**Vahedi-Langrudi**, 1996; **Karimi-Doostan**, 1997; **Karimi-Doostan**, 2005) and can be combined with
 108 different types of non-verbal items such as nominal, adjectivals or prepositional phrases (**Dabir-**
 109 **Moghaddam**, 1997).

110 In our study, we used separable complex predicates as defined by **Karimi-Doostan** (2011). According to
 111 **Karimi-Doostan**, a complex predicate can be separated if it satisfies both of the following two conditions:
 112 (1) if the nominal part is a noun to which adjectives, demonstratives, and wh-words, etc. can be attributed,
 113 and (2) if this noun has an internal argument structure (referring to an action or event). From this
 114 perspective, Persian complex predicates are categorized in three groups: (1) predicative verbal nouns (e.g.

115 *anja:m da:dan*, perform+to give), (2) predicative nouns (e.g. *latme zadan*, damage+to hit), and (3) non-
 116 predicative nouns (e.g. *gush da:dan*, ear+to do). Among these three types, only the second one satisfies
 117 both of the conditions.

118 We began by independently validating our assumption that the CPs we used in our experiments are
 119 predictable and separable. We first conducted a norming study (a sentence completion task), to establish
 120 that the light verbs (of the separated CPs) are highly predictable when the nominal is provided, as
 121 compared to non-CP verbs in simple predicate conditions. We then conducted an acceptability rating
 122 study to determine how acceptable Persian CPs are when they get separated.

3 NORMING STUDIES

123 In order to prepare appropriate stimuli, two norming studies were run. The first study involved sentence
 124 completion and served to validate (i) whether the identity of the verb in the complex predicate is highly
 125 predictable, and (ii) whether the identity of the verb in the control condition is not predictable.

126 The second study involved acceptability rating; the goal was to choose complex predicates for our
 127 experiments which are separable. That is, we wanted to identify complex predicates which native speakers
 128 would find acceptable even if an intervener occurs between the noun-verb sequence.

129 The sentence completion study was carried out to derive the predictions of the expectation account.
 130 Previous work on expectation effects suggests that sentence completion data may be useful for this
 131 purpose. For example, **Levy and Keller** (2013) used sentence completion data to complement their corpus
 132 analyses for deriving their predictions. In their study, the key issue was whether the intervening material
 133 (e.g., a dative marked NP) leads to a prediction of a dative verb. Their Table 4 shows that the intervening
 134 material sharpened the expectation for the type of verb predicted. This shows that sentence completion
 135 data can be used to determine empirically whether the prediction for a specific verb or a verb type is
 136 sharpened by intervening material; in the Levy and Keller case, it makes sense that the intervener sharpens
 137 the expectation, but clearly the nature and content of the intervening phrases will be crucial in determining
 138 whether expectations are sharpened (**Konieczny**, 2000; **Grodner and Gibson**, 2005).² Similarly, **Husain**
 139 **et al.** (2014) used sentence completion to establish that the identity of the verb in a complex predicate
 140 is highly predictable given the preceding context, but the identity of the verb in a simple predicate is
 141 not (see their Table 4). A third example is **Jäger et al.** (2015); they used both corpus data and sentence
 142 completion to establish that a sentence starting with a determiner, classifier, and an adverb leads to the
 143 prediction of a relative clause continuation in Chinese, and that the conditional probability of a subject
 144 relative continuation is higher than that of an object relative continuation (see their Table 2). Given these
 145 previous results, we assume that sentence completion data is informative about the predictions of the
 146 expectation-based account.

3.1 SENTENCE COMPLETION STUDIES

147 Two groups (32 participants each) of Persian native speakers, who did not take part in any of the other
 148 experiments, participated in two sentence completion pre-tests in which they were asked to complete the
 149 sentences after they were presented the sentence fragment until the pre-critical word. For example, as
 150 shown in 3, subjects were shown incomplete sentences which they had to complete; in this example, the
 151 missing verb is shown in parentheses. The participants were allowed to complete the sentence with as
 152 many words as they wanted, but our interest was only in the first word that they would write, which would
 153 most likely be a verb. This allowed us to calculate the proportion of continuations in which the exact verb
 154 was produced.

² We return to this point in the General Discussion, where we discuss the effect of entropy on reading times.

- 155 (3) a. Ali a:rezouyee bara:ye man (kard) ...
 156 Ali wish-INDEF for 1.S (do-PST) ...
 157 'Ali (made) a wish for me ...'
- 158 b. Ali a:rezouyee ke besya:r doost-da:sht-am bara:ye man (kard) ...
 159 Ali wish-INDEF that a lot like-1.S-PST for 1.S (do-PST) ...
 160 'Ali (made) a wish that I liked a lot for me ...'

161 The materials were exactly the same as the ones used in the experiments presented below. For experiment
 162 1 items, the average prediction accuracy for the exact verb in the strong predictability conditions was 76%
 163 for the short condition and 74% for the long condition; for experiment 2 items it was 77% and 76% for
 164 the short and long conditions respectively. By contrast, the average prediction accuracy for the exact verb
 165 in the weak predictability conditions in experiment 1 was 22% and 20% for the short and long conditions;
 166 and in experiment 2 it was 19% and 22% for the short and long conditions. An analysis using generalized
 167 linear mixed models shows a main effect of predictability in both the first experiment (coef = -1.32, SE =
 168 0.07, $z = -16.93$) as well as the second experiment (coef = -1.46, SE = 0.08, $z = -17.25$). As is clear from
 169 the mean percentages for each condition, the light verbs used in the complex predicate conditions were
 170 highly predictable, and the heavy verbs used in the simple predicate conditions were highly unpredictable.
 171 It is also clear from this study that, in our materials, increasing the amount of intervening material does
 172 not render the upcoming verb more predictable. The additional information provided by the intervening
 173 material for predicting the upcoming verb has been suggested by **Konieczny** (2000) as one possible
 174 explanation for shorter reading times at the verb in long- vs short-distance conditions. Although this
 175 proposal is likely to be correct for some constructions (see discussion in **Grodner and Gibson** 2005), in
 176 our materials, the sentence completion data do not provide any evidence that the intervening words we
 177 used in our design sharpen the expectation for the verb.³

3.2 ACCEPTABILITY RATING OF SEPARABLE VS INSEPARABLE CPS

178 Because the noun-verb sequences must be separable for our design to work, we also carried out an
 179 acceptability rating pre-test to make sure that the separability of the complex predicates used in our study is
 180 acceptable to native speakers. We tested for the acceptability of different types of noun-verb dependencies
 181 by interposing a short prepositional phrase between them. Taking Karimi-Doostan's classification of
 182 complex predicates into account, 36 items from each of the three categories were selected and randomized
 183 to test 50 native speakers of Persian (these participants did not take part in any other experiments reported
 184 here). They were asked to rate the sentences from 1 (unacceptable) to 7 (completely acceptable). Every
 185 subject saw all items. The average acceptability ratings for predicative verbal nouns, predicative nouns
 186 and non-predicative nouns were 3.23 (first quartile 1, third quartile 5), 6.08 (first quartile 6, third quartile
 187 7), and 3.12 (first quartile 1, third quartile 5) respectively. That is, items with predicative nouns were the
 188 most acceptable. We used all the 36 items of the predicative noun condition in our experiments 1, 2, and
 189 32 items in experiments 3, 4 (see the Methods section of experiment 3 for an explanation).

4 EXPERIMENT 1

4.1 METHOD

190 4.1.1 *Participants* Forty-two subjects aged between 17-40 years old (mean 24 years) participated in
 191 this experiment in Tehran, Iran. All participants were native speakers of Persian and were unaware of the

³ In fact, in our sentence completion data, as discussed in the General Discussion, entropy increases with distance.

192 purpose of the study. This study was carried out in accordance with the Helsinki Declaration, and letters
193 of consent were obtained from all the participants.

194 *4.1.2 Materials* We created 36 experimental sentences with a 2×2 factorial design, manipulating
195 predictability strength and distance between the object noun and verb. The short intervener was a
196 prepositional phrase and the long intervener was a relative clause added before the prepositional phrase.
197 In order to mask the experiment, we included 100 filler sentences with varying syntactic structures (see
198 Supplementary materials). Here is an example of the target sentences:

- 199 (4) a. Strong predictability, short distance (PP)
200 Ali a:rezouyee bara:ye man kard va...
Ali wish-INDEF for 1.S do-PST and...
201 ‘Ali made a wish for me and...’
202
- 203 b. Strong predictability, long distance (RC+PP)
204 Ali a:rezouyee ke besya:r doost-da:sht-am bara:ye man kard va...
Ali wish-INDEF that a lot like-1.S-PST for 1.S do-PST and...
205 ‘Ali made a wish that I liked a lot for me and...’
206
- 207 c. Weak predictability, short distance (PP)
208 Ali shokola:ti bara:ye man xarid va...
Ali chocolate-INDEF for 1.S buy-PST and...
209 ‘Ali bought a chocolate for me and...’
210
- 211 d. Weak predictability, long distance (RC+PP)
212 Ali shokola:ti ke besya:r doost-da:sht-am bara:ye man xarid va...
Ali chocolate-INDEF that a lot like-1.S-PST for 1.S buy-PST and...
213 ‘Ali bought a chocolate that I liked a lot for me and...’

214 The critical region is the verb (*kard* and *xarid*).

215 *4.1.3 Procedure* Participants were tested individually using a PC. They were explained the task before
216 they performed the self-paced reading (SPR) experiment. The participants were instructed to read for
217 comprehension in a normal manner and had a practice session of five sentences. All the sentences
218 were displayed on a single line and were presented in 22 pt Persian Arial font using Linger software
219 (<http://tedlab.mit.edu/~dr/Linger/>). In order to read each word of a sentence successively in a moving
220 window display, participants had to press the space bar; then the word seen previously was masked and
221 the next word was shown. After each sentence, they were asked to answer a comprehension question to
222 ensure that the participants paid attention to the complete sentence.

223 *4.1.4 Data analysis* The data analysis was conducted in the R programming environment (**R**
224 **Development Core Team**, 2013), using linear mixed-effects models (LMMs; **Pinheiro and Bates**
225 **2000; Bates et al.** 2015). For large samples, the t-distribution approximates the normal distribution
226 and an absolute value of t larger than 2 indicates a statistically significant effect at $\alpha = 0.05$. Sum
227 contrasts were used to code main effects and interactions. In addition, a nested contrast was defined
228 for a secondary analysis in order to look at the effect of distance in complex predicates vs the control
229 conditions separately; these were also coded as sum contrasts. For the reading time data, the most
230 complex model possible given the data and the design was chosen based on the `rePCA` function

231 (Bates et al., 2015); see the package `RePsychLing` (<https://github.com/dmbates/RePsychLing>) for
232 examples and more theoretical background. The `rePCA` function computes a principal components
233 analysis of the variance covariance matrices for the random effects (subject and item), which allows
234 the modeler to decide which variance components should be included. No attempt was made to fit
235 correlations between intercepts and slopes, for subjects or for items. All data and code are available
236 from <http://www.ling.uni-potsdam.de/~vasishth/code/SafaviEtAl2016DataCode.zip>.

4.2 PREDICTIONS (EXPERIMENT 1)

237 Based on the Husain et al. (2014) results, in experiment 1, we expected that increasing noun-verb distance
238 would lead to faster reading time at the verb in the strong predictable conditions, but slower reading time
239 in the weak predictable conditions. Thus, we expected to obtain a cross-over interaction.

240 The memory based accounts (Just and Carpenter, 1992; Gibson, 2000; Lewis and Vasishth, 2005)
241 predict that increasing distance should lead to a slowdown at the verb; these accounts make no predictions
242 about the strength of predictability.

243 There are two alternative predictions possible for the expectation account, depending on how one
244 operationalizes expectation. First, if sentence completion probabilities are a reasonable proxy for
245 conditional probabilities—and the previous research reported above (Husain et al., 2014; Levy and
246 Keller, 2013; Jäger et al., 2015) suggests that they may be—then we predict (a) no difference in reading
247 time at the verb as a function of distance, and (b) faster reading time at the verb in the strong predictable
248 conditions than the weak predictable conditions. Prediction (a) arises because, in the sentence completion
249 data, we see no effect of distance on the predictability of the upcoming verb, in either the strong or weak
250 predictability conditions; prediction (b) arises due to the difference in predictability of the exact verb that
251 we see in the strong versus weak predictability conditions (see the results of the sentence completion
252 studies).

253 An alternative possible prediction of the expectation account is that increasing distance should facilitate
254 processing at the verb. Surprisal predicts facilitation with increasing distance whenever distance causes
255 the number of possible parses to decrease; this decrease in the number of possible parses leads to the
256 probability mass being reassigned among the remaining parses. In our materials, when the participant
257 reads the noun in the noun-verb complex predicate, they are expecting the light verb with high probability
258 (nearly 1). However, in the long distance condition, the next word begins a relative clause; this leads to an
259 expectation that the light verb will appear *after* the relative clause verb. But what appears after the relative
260 clause verb is a PP that modifies the upcoming light verb. For a facilitation to be predicted in this long-
261 distance condition by the surprisal metric, it would have to be the case that the conditional probability of
262 the light verb following the RC and PP would be higher than the conditional probability of the light verb
263 in the short-distance (PP) condition. Whether this is true depends on what the facts are about Persian;
264 these are difficult to verify using corpus data because there would not be enough cases in the corpus of the
265 specific construction we investigate. It is possible that, if enough data were available, corpus counts could
266 in principle show that increasing distance leads to no difference in the conditional probability of a verb
267 appearing in the short vs long conditions; this would be true if the corpus-based conditional probability
268 of the verb coming up remains unchanged in both the short and long conditions. In order to get a sense
269 of how the conditional probabilities change in the noun-light verb condition as a function of distance,
270 we extracted all light verb sentences from a Persian corpus (Seraji, 2015) and then counted, for different
271 numbers of modifying phrases, the proportion of cases that a verb followed the intervening phrase. For
272 example, in a Persian sentence such as *John in the morning went*, there is one intervening phrase, the PP.
273 As shown in Table 1, we find that the conditional probability of the verb appearing next is always high,
274 but goes to 1 with increasing distance. This suggests that in general, increasing distance tends to sharpen
275 the expectation for an upcoming verb. Of course, these corpus counts don't give us any direct information
276 about the predictions regarding our particular experiment design.

Table 1. The conditional probability of a light verb appearing given the complex predicate noun and n intervening phrases between the noun and the light verb.

n intervening phrases	probability of verb
0	$3826/4003 = 0.95$
1	$131/133 = 0.98$
2	$28/31 = 0.90$
3	$5/5 = 1$
4	$2/2 = 1$
6	$1/1 = 1$

277 Regarding the strong vs weak predictability conditions, note that the expectation account of Hale and
 278 Levy does not predict that processing should be facilitated when the exact identity of the upcoming verb is
 279 predicted (strong predictability condition), compared to the case when just some verb is predicted (weak
 280 predictability condition). This is because the surprisal metric is usually calculated using the conditional
 281 probability of the part-of-speech (verb) given preceding context, and this will be the same in both the
 282 strong and weak predictability conditions. However, it is possible to subsume the difference between
 283 strong and weak predictability under the surprisal account by reframing the conditional probabilities in
 284 terms of the exact identity of the verb. In this case, the expectation account would predict faster reading
 285 times in the strong predictability conditions compared to the weak predictability conditions, regardless of
 286 distance.

287 To summarize, regarding the distance manipulation, the expectation account predicts either no effect
 288 or a facilitation at the verb as a function of distance; and regarding the predictability manipulation, the
 289 expectation account (appropriately formulated to include the conditional probability of the exact lexical
 290 item predicted) would predict a main effect of predictability.

4.3 RESULTS

291 *4.3.1 Comprehension accuracy* Participants answered correctly on average 92.73 percent of all
 292 comprehension questions (excluding fillers). Accuracy was 91, 94, 95 and 91 percent respectively for the
 293 four conditions in (1). A generalized linear mixed model of the binary responses showed an interaction
 294 (coef=-0.25, SE=0.10, $z = -2.37$) between predictability and distance. A nested contrast suggests that this
 295 interaction is driven by the weak predictability condition, such that response accuracy is lower in the long
 296 condition compared to the short condition.

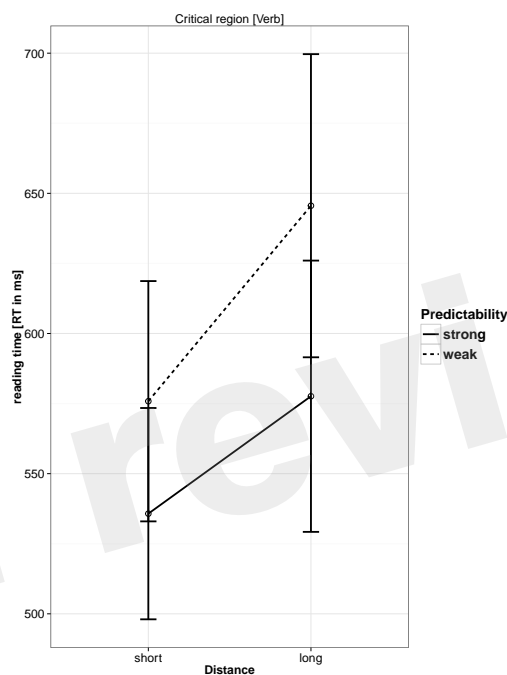
297 *4.3.2 Reading time* Reading times (RTs) were analyzed at the verb; plots of the other regions are shown
 298 in the Supplementary materials. As shown in Table 2 and Figure 1, there was a main effect of distance,
 299 such that increasing distance led to longer reading times. There was also a main effect of predictability: the
 300 complex predicate conditions were read faster overall. A marginal interaction is also seen: stronger locality
 301 effects are seen in the control condition than in the complex predicate condition. A nested analysis shows
 302 that the distance effect was driven by the control (weak predictability) condition (strong predictability:
 303 coef.=0.02, SE=0.015, $t=1.53$; weak predictability: coef.=0.06, SE=0.015, $t=3.87$).

4.4 DISCUSSION

304 Experiment 1 found a main effect of predictability such that the strong predictability conditions were
 305 read faster than the weak predictability conditions, and a main effect of distance, such that the short
 306 conditions were read faster than the long conditions. A nested contrast showed that this effect of distance
 307 was driven by the weak predictability conditions, i.e., reading time at the verb in condition c was faster

Table 2. Coefficients, standard errors, and t-values for the main effects and interactions in Experiment 1.

Comparison	Coefficient	SE	t-value
(Intercept)	6.24	0.04	151.99
Distance	0.04	0.01	3.88
Predictability	-0.03	0.01	-2.94
Distance x Predictability	0.02	0.01	1.70

**Figure 1.** Reading times at the critical verb in Experiment 1.

308 than the reading time in condition d. A marginal interaction suggests that the locality effect may be
 309 somewhat stronger in the weak predictability condition. The marginal interaction seems to provide only
 310 weak support for the idea that strong predictability can at least attenuate locality effects (Husain et al.,
 311 2014). The results are partly consistent with memory-based accounts, which correctly predict a slowdown
 312 at the verb in the long conditions, i.e., a main effect of distance. However, as the nested comparison
 313 shows, the main effect of distance is driven only by the weak predictability (non-complex predicate)
 314 conditions. Memory-based theories would be unable to explain this because they predict a slowdown
 315 in long conditions irrespective of predictability strength. The expectation account's prediction regarding
 316 distance, that increasing the argument-verb distance would either have no effect or result in a facilitation,
 317 was not validated; however, the main effect of predictability is consistent with a version of the expectation
 318 account that uses the conditional probability of the exact lexical item (verb) appearing given the preceding
 319 context.

320 Our original motivation for this study was to attempt a replication of the Husain et al. (2014) findings.
 321 The results are not entirely inconsistent with those of Husain et al. (2014), but they are also not a strong
 322 validation of the expectation-memory cost tradeoff posited in that paper. As in the Husain et al. study,
 323 we see a main effect of predictability driven by the complex predicate condition. This effect could be

324 explained in terms of reduced retrieval cost at the verb due to its high expectation. An obvious confounding
 325 factor here is that the verbs in the strong vs weak predictability conditions are not identical; this prevents
 326 us from ruling out the possibility that low-level differences in the verbs might be responsible for the
 327 facilitation due to prediction strength.

328 We turn next to experiment 2, in which we manipulate the type of intervener. Here, in the long distance
 329 condition, instead of a relative clause and PP intervener, a long PP intervenes. The motivation was to
 330 increase distance without having different types of interveners in the short vs long conditions, as this
 331 might be a fairer comparison.

5 EXPERIMENT 2

5.1 METHOD

332 *5.1.1 Participants* Forty-three subjects, with the same criteria as in experiment 1, participated in this
 333 experiment in Tehran, Iran. This study was carried out in accordance with Helsinki Declaration, and
 334 consent forms were obtained from all the participants.

335 *5.1.2 Materials* The stimuli and fillers were the same as in experiment 1 except, for the long conditions
 336 (b and d) where the intervener was a longer prepositional phrase (PP) instead of the combination of a
 337 relative clause and a PP as in the previous experiment. The PP was lengthened using several different
 338 structures, all of which had one or more instance of the ezafe possessive marker (**Samvelian, 2007**):

- 339 1. N-ez N-ez N/pronoun/proper name
- 340 2. N-ez adj-ez N/pronoun/proper name
- 341 3. N-ez adj-ez N
- 342 4. N-ez N-ez adj
- 343 5. N adj-ez adj
- 344 6. superlative adj N N/pronoun/proper name
- 345 7. N-ez pronoun

346 One set of examples using the first type of PP shown above is as follows :

- 347 (5) a. Strong predictability, short distance (PP)
 348 Ali a:rezouyee bara:ye man kard va...
 Ali wish-INDEF for 1.S do-PST and...
 349 'Ali made a wish for me and...'
 350
- b. Strong predictability, long distance (longer PP)
 352 Ali a:rezouyee bara:ye doost-e xa:har-e man kard va...
 Ali wish-INDEF for friend-EZ sister-EZ 1.S do-PST and...
 353 'Ali made a wish for my sister's friend...'
 354
- c. Weak predictability, short distance (PP)
 355 Ali shokola:ti bara:ye man xarid va...
 356 Ali chocolate-INDEF for 1.S buy-PST and...

357 'Ali bought a chocolate for me and...'

358

359 d. weak predictability, long distance (longer PP)

360 Ali shokola:ti bara:ye doost-e xa:har-e man xarid va...
 Ali chocolate-INDEF for friend-EZ sister-EZ 1.S buy-PST and...

361 'Ali bought a chocolate for my sister's friend and...'

362

363 More details about the PPs are provided in the Supplementary materials.

364 5.1.3 *Procedure and Data Analysis* The procedure and data analysis methodology was the same as
 365 experiment 1.

5.2 PREDICTIONS (EXPERIMENT 2)

366 In experiment 2, since the distance manipulation involves lengthening the PP, surprisal will predict no
 367 difference at the verb. This is because the end of the PP raises a strong expectation for a verb, and this
 368 strong expectation for a verb should be the same in both the short and long PP conditions.

5.3 RESULTS

369 5.3.1 *Comprehension Accuracy* Participants answered 92.75 percent of all comprehension questions
 370 correctly on average (excluding fillers). The accuracies by condition were 96, 92, 94, and 89 percent
 371 respectively for the four conditions in (2). The generalized linear mixed models of the responses showed
 372 a main effect of distance (coef=-0.35, SE=0.10, z=-3.39) such that accuracies were lower in the long
 373 conditions. No effect of predictability strength, and no interaction between predictability strength and
 374 distance were found.

375 5.3.2 *Reading Time* As shown in Table 3 and Figure 2, the results showed a main effect of distance,
 376 with long distance conditions being read slower. There was also an effect of predictability, with the strong
 377 predictability condition being read faster than the weak predictability condition. No interaction was found
 378 between predictability and distance. A nested contrast showed that the distance effect is seen in both
 379 strong predictability (coef.=0.06, SE=0.02, t=3.63) and weak predictability (coef.=0.05, SE=0.02, t=2.62)
 380 conditions.

Table 3. Coefficients, standard errors, and t-values for the main effects and interactions in Experiment 2.

Comparison	Coefficient	SE	t-value
Intercept	6.27	0.04	147.86
Distance	0.06	0.01	3.99
Predictability	-0.02	0.01	-2.28
Distance x Predictability	-0.01	0.01	-0.67

5.4 DISCUSSION

381 In this experiment, we replicated the locality effects found in Experiment 1, but we no longer see a
 382 weakening of the locality effect that was seen in Experiment 1 (a marginal interaction was found in

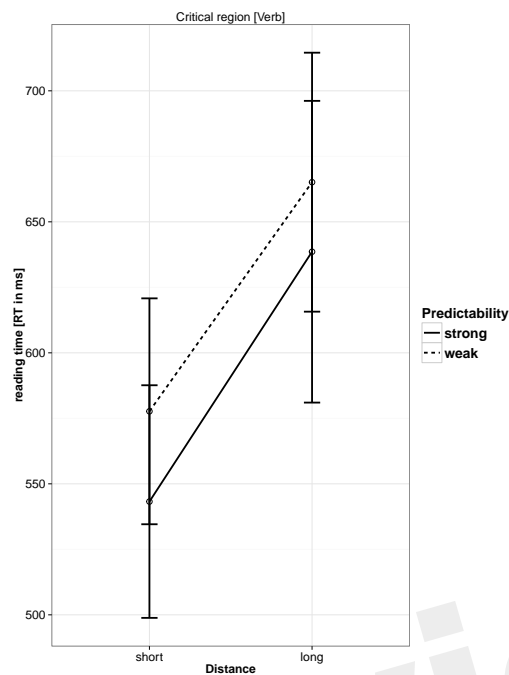


Figure 2. Reading times at the critical verb in experiment 2.

383 Experiment 1). Nested contrasts showed that locality effects are equally strong in both the strong and
 384 weak predictability conditions. In Experiment 2, we also see an effect of predictability, with the strong
 385 predictable verb being read faster. Thus, regarding the distance manipulation, the working-memory
 386 account's prediction is validated, and the expectation-based account's prediction is not supported. The
 387 main effect of predictability does furnish evidence consistent with the expectation-based account.

388 A secondary analysis was conducted to compare the strength of the locality effect in the two experiments,
 389 and to determine whether the interaction between distance, predictability and experiment is significant.
 390 The between-subject factor experiment was coded using sum coding: experiment 1 was coded -1, and
 391 experiment 2 was coded +1 (further details are available in the Supplementary materials). The results are
 392 shown in Table 4. An interaction between distance and experiment is seen: the locality effect was stronger
 393 in experiment 2. This confirms the suggestion that the locality effect is strengthened in experiment 2
 394 compared to experiment 1.

Table 4. Comparison of experiments 1 and 2.

Comparison	Coef.	SE	t-value
Intercept	6.23	0.03	231.29
Distance	0.03	0.01	3.72
Predictability	-0.02	0.01	-2.25
Expt	0.04	0.03	1.67
Distance × Predictability	-0.00	0.01	-0.55
Distance × Expt	0.03	0.01	5.58
Predictability × Expt	-0.01	0.01	-1.27
Pred × Dist × Expt	0.01	0.01	1.66

395 In experiment 2, the intervener was a long, uninterrupted prepositional phrase whereas in experiment
396 1, the intervener consisted of a short RC followed by a PP. One can speculate about why experiment 2
397 shows equally strong effects in both predictability conditions: Processing a single long intervening phrase
398 may be harder than processing two different phrases because it may be harder to chunk a single long
399 phrase compared to two shorter phrases; this is predicted by the Sausage Machine proposal of **Frazier**
400 **and Fodor** (1978). If this is correct, then the complexity of the intervener may indeed be a relevant factor
401 in determining whether strong expectation can weaken locality effects. It is possible to test this claim
402 by using an intervener that is much easier to process; an example would be an adverb containing no noun
403 phrases.

404 We were motivated by the recent replication crisis in psychology (**Open Science Collaboration**, 2015)
405 to attempt to replicate our results using a different method. Furthermore, replications using eye-tracking
406 would be very informative because it is possible that self-paced reading overburdens the working-memory
407 system in an unnatural manner. If this is the case, one prediction would be that the eye-tracking data would
408 not necessarily show locality effects. We describe these experiments next.

6 EXPERIMENT 3

6.1 METHOD

409 *6.1.1 Participants* Forty subjects, with the same criteria for inclusion as in the previous experiments,
410 participated in the eye-tracking study in University of Potsdam, Germany.

411 *6.1.2 Materials* The experimental items were exactly the same as experiment 1 (self-paced reading),
412 except that four items from experiment 1 were removed. The following four items were removed: item
413 id 5, *sheka:yat kardan* (complain + to do), item id 9, *sahm bordan* (share + to win), item id 26, *pishraft*
414 *kardan* (progress + to do), and item id 32, *hes kardan* (feel + to do). The reason was that the results
415 of the sentence-completion studies suggested that these light verbs had lower predictability than the
416 other light verbs in the stimuli. It could be that this lower predictability is due to the existence of some
417 other alternative light verbs with which the nominal part can combine to make other possible complex
418 predicates. The last two CPs also had a lower acceptability rating (item 26 had 4.7, and item 32 had 3.5).
419 As a consequence, in our eye-tracking study, we had thirty-two experimental items and sixty-four fillers.
420 All items, including fillers are available in the Supplementary materials.⁴

421 *6.1.3 Procedure* An eye-tracking study was prepared using Experiment-Builder software, and
422 participants' eye-movements were recorded using an EyeLink 1000 tracker, with a connection to a
423 PC. Before the experiment started, the participants were instructed to read the sentences silently at a
424 normal pace and had a practice block consisting of five sentences. After answering the comprehension
425 questions of the practice block, they were provided with feedback indicating whether or not the answer
426 was correct. A 21-inch monitor was placed 60 centimeters from the participants' eyes. In order to reduce
427 head movements, the participants were asked to use the chin-rest. They viewed the sentences with both
428 eyes, but only the right eye was recorded. The items were presented in one line and in 18 points Persian
429 Arial font (from right to left). First, they had to fixate on a dot at the right edge of the screen so that the
430 sentence appeared. After they finished reading, they had to fixate on the dot in the bottom left corner of
431 the screen; once they fixated on the dot, the comprehension question was presented. Unlike the practice
432 items, they were not provided with any feedback. Calibration was performed at the beginning of the
433 experiment, after their 5-minute break (which occurred after they had read half the items), and whenever
434 it was necessary.

⁴ We also reanalyzed Experiments 1 and 2 after removing these four items; this did not change the results reported above for the experiments.

435 6.1.4 *Data analysis* Raw gaze duration data was obtained using the Data Viewer software.⁵ This data
 436 was then processed to get different eye-tracking measures using em2 (Logačev and Vasishth, 2014). As
 437 discussed earlier, linear mixed models were used for the analysis. All analyses were carried out using
 438 log-transformed data. For each reading time measure, zero ms reading times were removed.

6.2 RESULTS

439 6.2.1 *Comprehension accuracy* On average, participants correctly answered 91.05 percent of the target
 440 comprehension questions. Also, per condition, they had 91 percent response accuracy for condition a, 91
 441 percent for condition b, 95 percent for condition c, and 89 percent for condition d. The generalized
 442 linear mixed models of the responses showed a main effect of distance (coef=-0.26, SE= 0.11, z= -2.39,
 443 p=0.016). An interaction was also found between predictability and distance (coef=-0.26, SE= 0.11, z=-
 444 2.38, p=0.016), and the nested analysis suggested that the interaction derives from the weak predictability
 445 conditions showing lower accuracies in the long vs short distance conditions (coef=-0.52, SE=0.16, z=-
 446 3.14, p=0.001).

447 6.2.2 *Reading time* The critical region was the verb, as in Experiments 1 and 2. The same sum contrast
 448 coding was used as in experiments 1 and 2; in addition, nested contrast coding was used to investigate the
 449 effect of distance within the two predictability conditions. We present results for first-pass reading time,
 450 regression path duration, and total reading time.

451 The effect of predictability, seen in Experiments 1 and 2, is also present in first-pass reading time
 452 and total reading time; the strong-predictability conditions had shorter reading times. Also, as in
 453 Experiments 1 and 2, there was an effect of distance; the long-distance conditions has longer reading
 454 times. Table 6 shows the details of the analyses. A nested contrast showed that in first-pass reading time the
 455 distance effect was present in both the strong- and weak-predictability conditions (strong predictability:
 456 coef.=0.044, SE=0.02, t=2.19; weak predictability: coef.=0.06, SE=0.02, t=2.57). Regression path
 457 duration did not show any distance effects with the two predictability conditions (strong predictability:
 458 coef.=0.03, SE=0.03, t=1.23; weak predictability: coef.=0.04, SE=0.03, t=1.49). The nested contrast in
 459 total reading time showed no effect of distance in the strong-predictability condition (coef.=0.03, SE=0.03,
 460 t=1.27), but a distance effect was seen in the weak predictability condition (coef.=0.07, SE=0.026, t=2.52).

Table 5. Coefficients, standard errors, and t-values for the main effects and interactions in Experiment 3.

ET 1 measures	Comparison	Coef	SE	t value
Log FPRT	Intercept	5.62	0.03	175.88
	Distance	0.05	0.02	2.67
	Predictability	-0.053	0.02	-3.08
Log RPD	Distance × Predictability	0.01	0.01	0.79
	Intercept	5.73	0.04	128.98
	Distance	0.04	0.02	1.47
Log TRT	Predictability	-0.08	0.02	-3.21
	Distance × Predictability	0.004	0.02	0.26
	Intercept	5.77	0.05	124.04
Log TRT	Distance	0.05	0.02	2.17
	Predictability	-0.10	0.02	-4.17
	Distance × Predictability	0.02	0.02	0.99

⁵ <http://www.sr-research.com/dv.html>

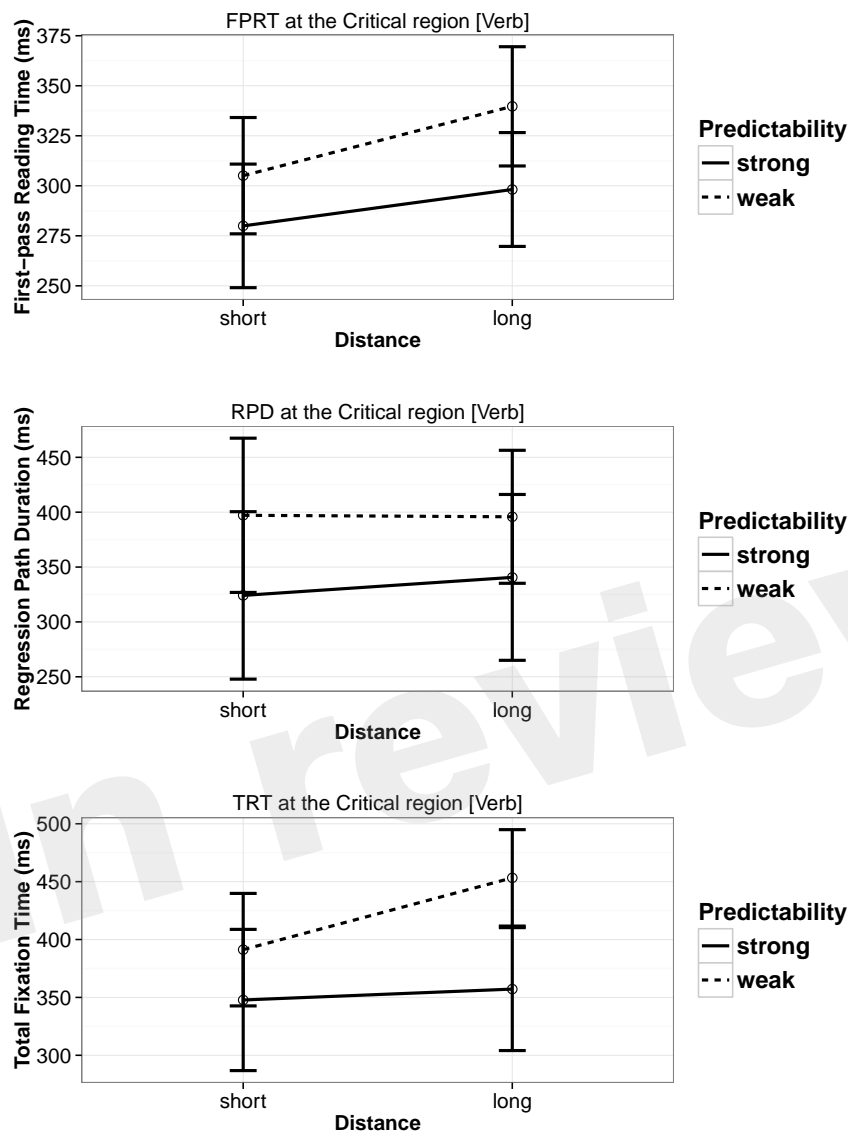


Figure 3. First-pass reading time, regression path duration, and total reading time in Experiment 3 at the critical verb.

6.3 DISCUSSION

461 In the eye-tracking Experiment 3, we replicated the locality effects found in the Experiment 1 in first-pass
 462 reading time and total reading time. Nested contrasts showed that the locality effect tends to appear in
 463 weak-predictability conditions, which is similar to the result in Experiment 1. In first-pass reading time,
 464 the locality effect appeared in both the strong and weak-predictability conditions, but the magnitude of
 465 the effect was stronger in the weak-predictability condition. A main effect of predictability was found in
 466 all three dependent measures, replicating the effect in Experiment 1.

467 Since we failed to find any interaction between predictability and distance, we cannot conclude,
 468 as **Husain et al.** (2014) did, that expectation effects can cancel locality effects. The locality effects
 469 are consistent with the working memory accounts (**Gibson, 2000; Lewis and Vasishth, 2005**) and

470 inconsistent with the distance-based predictions of the expectation account. As in the SPR experiments, we
471 have evidence consistent with a version of the expectation account that predicts that strong predictability
472 conditions will be read faster than the weak predictability conditions.

473 In the strong-predictability conditions, the somewhat weaker locality effect seen in first-pass reading
474 time, and the absence of the effect in total reading time could be taken to be weakly consistent with
475 the claims in **Husain et al.** (2014), but without an interaction between distance and predictability, these
476 patterns are not really convincing.

477 In sum, the main result in experiment 3 is that we have replicated the locality effect and the facilitation
478 due to strong predictability. There is some weak evidence that the locality effect may be reduced in the
479 strong-predictability condition; but the absence of an interaction does not support the claim in **Husain**
480 **et al.** (2014), that strong expectations cancel locality effects; the most we can say from the eye-tracking
481 data is that strong expectations may weaken locality effects.

7 EXPERIMENT 4

7.1 METHOD

482 *7.1.1 Participants* Forty participants, with the same criteria as in the previous experiments,
483 participated in the eye-tracking study in Golm campus, University of Potsdam, Germany.

484 *7.1.2 Materials* The experimental items were exactly the same as experiment 2 (self-paced reading),
485 but with 32 items (see the explanation for Experiment 3 regarding the four items that were removed).
486 The experimental items were complemented with 64 filler sentences with varying syntactic structures (see
487 Supplementary materials).

488 *7.1.3 Procedure and Data Analysis* The procedure and data analysis were exactly the same as
489 experiment 3 (eye-tracking).

7.2 RESULTS

490 *7.2.1 Comprehension Accuracy* On average, participants answered 90.05 percent of comprehension
491 questions correctly. They had 94 percent response accuracy for condition a, 88 percent for condition b,
492 94 percent for condition c, and 86 percent for condition d. The generalized linear mixed models of the
493 responses showed a main effect of prediction (coef=0.86, SE=0.40, $z=2.13$, $p=0.033$). Also there was an
494 interaction between predictability and distance (coef=-1.034, SE=0.50, $z=-2.04$, $p=0.041$), and the nested
495 analysis shows that this interaction derives from lower accuracy in the long vs short distance conditions
496 in the weak-predictability conditions (coef=-0.52, SE=0.19, $z=-2.65$, $p=0.007$).

497 *7.2.2 Eye-tracking measures* Unlike experiment 3, in the current experiment, we found effects of
498 distance and predictability in all the three measures (see Table 6). In other words, in the three measures
499 example, the long conditions (b and d) were read slower than the short conditions (a and c), and the weak
500 predictability conditions (c and d) were read slower than the strong predictability conditions (a and b).
501 None of the measures showed any interaction between predictability and distance.

502 Nested comparisons showed that in first-pass reading time, the locality effect was seen in the strong-
503 predictability condition (coef.=0.05, SE=0.02, $t=2.33$), but there was a weaker tendency towards a
504 locality effect in the low-predictability condition (coef.=0.06, SE=0.03, $t=1.86$). In regression-path
505 duration, both strong- and weak-predictability conditions showed a locality effect (strong-predictability:
506 coef.=0.086, SE=0.02, $t=3.57$; low-predictability: coef.=0.07, SE=0.03, $t=2.15$). In total reading time,
507 the strong-predictability condition showed a locality effect (coef.=0.10, SE=0.02, $t=4.06$), but in the

508 low-predictability condition, only a tendency towards a locality effect was seen (coef.=0.06, SE=0.04,
509 t=1.59).

Table 6. Coefficients, standard errors, and t-values for the main effects and interactions in Experiment 4.

ET measures	Comparison	Coef	SE	t value
Log FPRT	Intercept	5.67	0.04	147.60
	Distance	0.06	0.02	2.46
	Predictability	-0.11	0.02	-5.10
	Distance × Predictability	-0.01	0.02	-0.30
Log RPD	Intercept	5.79	0.05	125.45
	Distance	0.08	0.02	3.63
	Predictability	-0.11	0.02	-4.76
	Distance × Predictability	0.01	0.02	0.43
Log TRT	Intercept	5.80	0.05	122.09
	Distance	0.08	0.03	3.06
	Predictability	-0.15	0.02	-6.69
	Distance × Predictability	0.02	0.02	1.27

7.3 DISCUSSION

510 The eye-tracking Experiment 4 replicated the results of the self-paced reading study (Experiment 2): a
511 main effect of distance and a main effect of predictability, with no evidence for an interaction. The effects
512 in early (FPRT), regression (RPD), and late (TFT) measures showed the same patterns as in the first eye-
513 tracking study. However, the locality effects were even stronger, in the same way that the second self-paced
514 reading study showed stronger locality effects. Also, these effects are equally strong in both strong and
515 weak predictability conditions, mirroring our finding in the second self-paced reading study. Evidence that
516 the long PP in experiment 4 leads to stronger locality effects than in experiment 3, which has an RC+PP
517 intervener, comes from a combined analysis of the two eye-tracking experiments, including experiment
518 as a between subjects factor; this showed significant interactions between distance and experiment in
519 regression path duration. In other words, the locality effect was stronger in experiment 4 than experiment
520 3 (coef. 0.02, SE 0.01, t= 2.14).

521 Overall, regarding the distance manipulation, the results are consistent with memory-based accounts,
522 and inconsistent with the expectation account. The main effect of predictability is consistent with the
523 expectation account, as discussed earlier. In Experiment 4, we don't see any evidence consistent with
524 the **Husain et al.** (2014) proposal; if anything, the locality effect is *stronger* in the strong-predictability
525 conditions.

8 GENERAL DISCUSSION

8.1 EVALUATING THE PREDICTIONS OF THE MEMORY-BASED ACCOUNTS AND THE EXPECTATION-BASED ACCOUNT

526 As summarized graphically in Figure 5, Our main finding from the four Persian studies is that the
527 locality effect predicted by memory accounts is upheld, but there is no evidence for the expectation-
528 based account's prediction of facilitation in longer distance conditions. We consistently see a main effect
529 of predictability, which is consistent with expectation accounts. Finally, there is no compelling evidence
530 in the Persian data that strong expectations cancel locality effects.

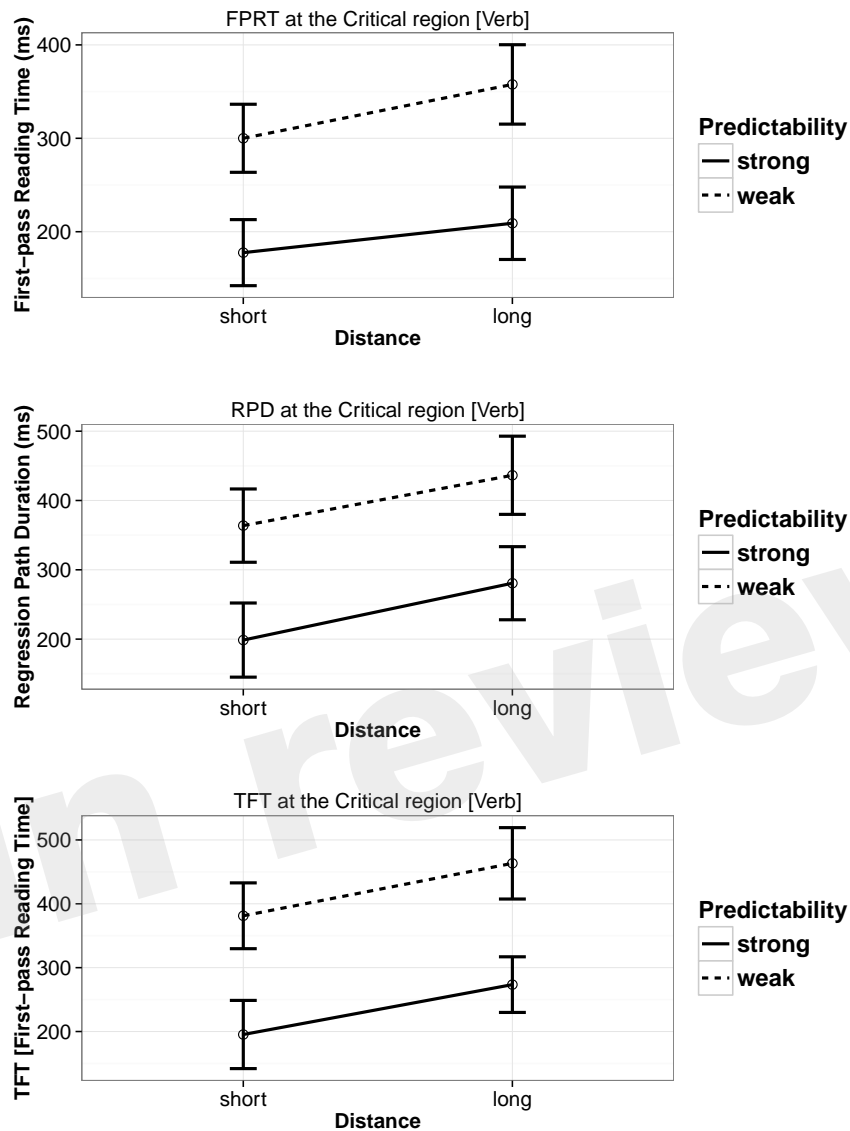


Figure 4. First-pass reading time, regression path duration, and total reading time in Experiment 4 at the critical verb.

531 There is also suggestive evidence that the complexity of intervening material could strengthen the
 532 locality effect: when the intervener is an RC followed by a PP, we see a marginal interaction between
 533 distance and predictability, but when the intervener is a single long PP, we see no evidence for an
 534 interaction between distance and predictability strength, and we tend to see stronger effects. The two
 535 combined analyses of the SPR studies and of the eye-tracking studies show that the locality effect is
 536 stronger in the experiments with the long PP. Of course, a definitive test of such a difference would be
 537 a new design where we directly compare intervener types in a within-subjects design; such a follow-up
 538 study is currently being planned.

539 We consistently found a main effect of predictability in all four experiments: the strong predictability
 540 conditions were read faster at the verb than the weak predictability conditions. This is consistent with
 541 the expectation-based account. Since the verbs in the strong and weak predictability conditions are

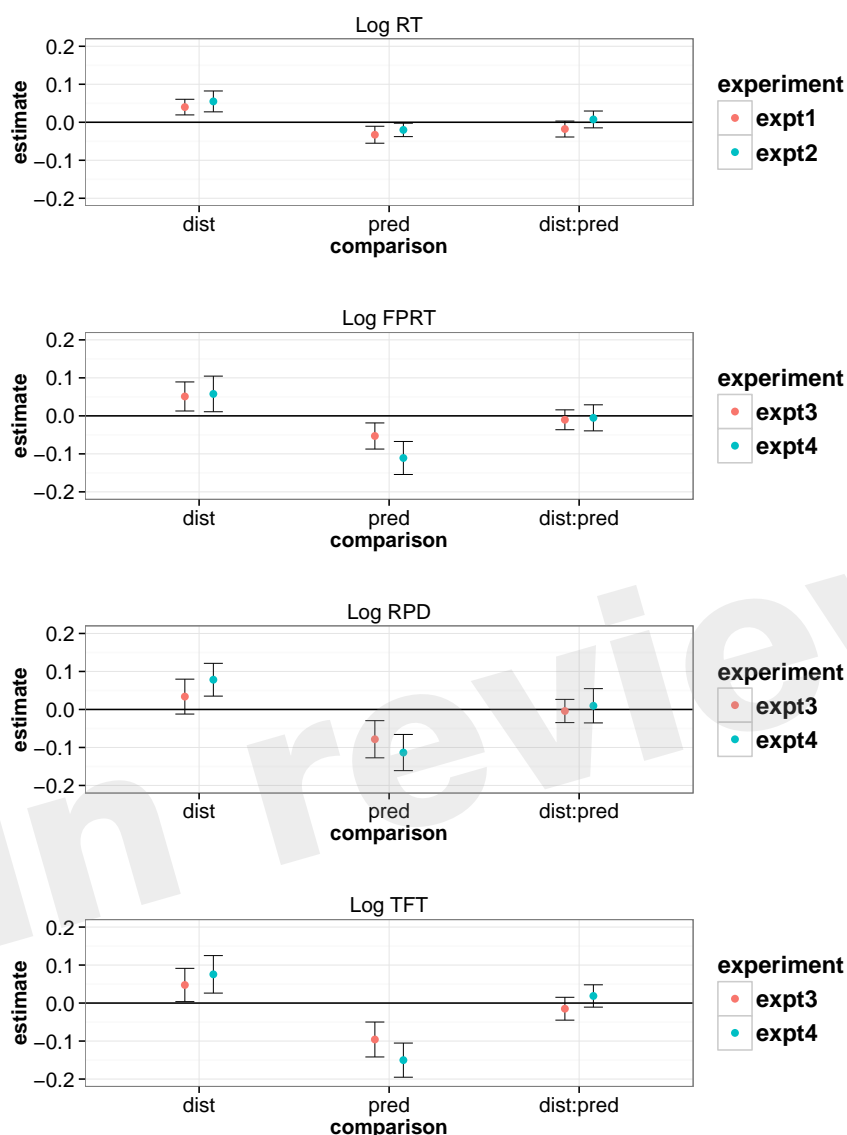


Figure 5. Summary of the magnitudes of effects (derived from the linear mixed models) across the four experiments. The error bars show 95% confidence intervals.

542 not identical, we cannot rule out the possibility that word frequency or other such low-level factors
 543 are responsible for these effects. However, it is plausible that the highly predictable verb is processed
 544 faster than the less predictable verb. Thus, the main effect of predictability can be seen as evidence for
 545 expectation-based accounts, operationalized in terms of the conditional probabilities of the appearance of
 546 the exact verb given the preceding context.

547 It is possible that we were unable to replicate the Husain et al findings because of the nature of the
 548 intervener used in the Persian studies. Unlike **Husain et al.** (2014) where the long distance condition had
 549 extra adverbials compared to the short condition, in Experiment 1 we have a more complex intervener,
 550 a relative clause. Another reason for finding the effects which are different from the study by **Husain**
 551 **et al.** (2014) could be that in Persian, separating the nominal part of the CP from the light verb occurs

552 relatively rarely, compared to Hindi. There is some support for this in corpus data. Based on the Hindi
 553 dependency treebank (Bhatt et al., 2009), the average distance, counted as the number of intervening
 554 phrases, between an object and its (heavy) verb in Hindi is 0.82 (with minimum 0 and maximum 15, and
 555 first and third quantiles 0 and 1), and the average distance between a noun and light verb is .07 (minimum
 556 0 and maximum 18, with first and third quantiles 0 and 0). In the Persian dependency treebank (Seraji,
 557 2015), the average distance between an object and (heavy) verb is 2.48 (with minimum 0 and maximum
 558 9, and first and third quantiles 1 and 3), while the average distance between a noun and light verb is
 559 0.05 (with minimum 0, and maximum 6, and first and third quantiles 0 and 0). Thus, the adjacency of
 560 CPs in Persian is strongly preferred (maximum 6 vs Hindi's maximum 18), although as validated in the
 561 acceptability rating norming study, this separability is acceptable and not considered ungrammatical.⁶

8.2 AN ALTERNATIVE EXPLANATION IN TERMS OF ENTROPY

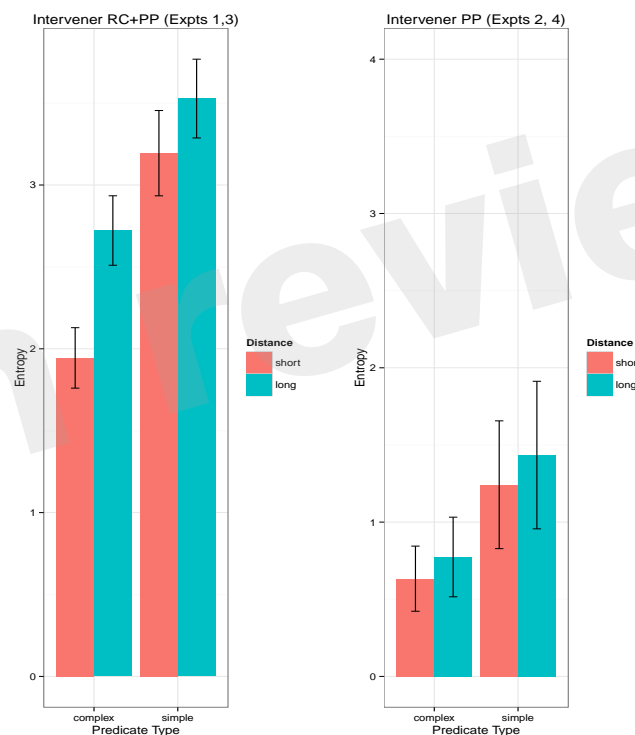


Figure 6. The estimated entropy (with 95% confidence intervals), computed using the sentence completion data, for the two experiment designs.

562 Could there be an alternative explanation for the locality effect seen in the four experiments, one
 563 that does not invoke greater memory cost in the long-distance conditions? One possibility is that
 564 entropy (uncertainty) increases with increasing distance. Entropy is an information-theoretic measure
 565 that essentially represents how uncertain we are of the outcome (Shannon, 2001). In the present case,
 566 this would translate to our uncertainty about the upcoming verb. If there are n possible ways to continue
 567 a sentence, and each of the possible ways has probability p_i , where $i = 1, \dots, n$, then entropy is defined

⁶ These intervening phrases have been computed using dependency treebanks. Consequently, phrasal boundaries are approximations. Also, because of annotation differences between the two treebanks, phrase boundary criteria sometimes differ for the two languages. The phrasal counts lead to the same conclusions regardless of whether one counts intervening phrases or words.

568 (Shannon, 2001) as $-\sum_i p_i \times \log_2(p_i)$. The entropy associated with the upcoming verb can be calculated
569 using our sentence completion data.⁷

570 In order to evaluate whether entropy could explain the locality data, we computed entropy for each item
571 in each condition for both experiments. The estimated entropies for each condition in the two experiment
572 designs are shown in Figure 6. It is important to note here that entropy for each condition in Figure 6
573 is based only on nine data points per condition (we only have $9 \times 4 = 36$ items); for different items,
574 there is substantial variability in the entropy patterns by condition. Nevertheless, in the figure we can see
575 that in the items used for Experiments 1 and 3, the entropy is higher in the long-distance conditions. The
576 effect of entropy is less clear for the items used in Experiments 2 and 4, because of the relatively wider
577 confidence intervals. Clearly uncertainty is higher in the RC+PP conditions. A closer look at the high
578 predictability conditions shows that the entropy difference between the long and short distance conditions
579 is larger in the RC+PP intervener items than the entropy difference in the long PP intervener items (it is
580 larger by 0.28, with 95% confidence intervals -0.01 and 0.57). This suggests that the intervening RC may
581 be responsible for creating a greater degree of uncertainty regarding the upcoming verb. Thus, the entropy
582 patterns by condition suggest that uncertainty may increase with argument-verb distance; especially in
583 the strong predictability conditions, increasing distance is not sharpening the expectation for the verb (cf.
584 Konieczny (2000)), and the RC intervener may be the cause for the greater entropy in the long-distance
585 conditions. The reason for the RC causing an increase in entropy needs further study; it would also be
586 useful to revisit the existing locality effects in English and Hindi from the perspective of entropy.

587 In order to investigate whether entropy affects reading times at the verb, we fit linear mixed models with
588 predicate type, distance, as sum-coded factors, and entropy (centered) as a continuous factor; all higher
589 order interactions were also included. Varying slopes for entropy were always included with varying
590 intercepts for item, and other varying slopes were also included when these were justified (Bates et al.,
591 2015); no attempt was made to estimate intercept-slope correlations. The dependent variable was log
592 reading time at the critical verb. In Experiment 1, in addition to the effects of predictability and distance,
593 we find an effect of entropy (coef.=0.05, SE=0.02, t=2.8), and an interaction between distance and entropy
594 (coef.=0.04, SE= 0.02, t= 2.3), such that long distance conditions lead to a greater effect of entropy. None
595 of the other experiments showed any effects of entropy. Thus, although the evidence in favor of entropy
596 is not overwhelming, a potentially important finding here is that entropy could explain locality effects.
597 To our knowledge, this is the first demonstration that locality effects may arise due to factors other than
598 memory costs. More research is needed to establish whether entropy can be a general explanation for
599 locality effects.

600 In conclusion, as regards the distance manipulation, the evidence from Persian is in favor of working-
601 memory accounts, although entropy is also a candidate explanation. There is not much evidence
602 from Persian that strong-predictability conditions cancel locality effects, as Husain and colleagues had
603 suggested. Interestingly, there is no evidence in these experiments for the prediction of the expectation
604 account regarding the distance manipulation, that increasing argument-verb distance facilitates processing
605 due to increasing conditional probabilities of the upcoming verb. The suggestion in (Levy et al., 2013)
606 that “the verb-medial languages tend to exhibit the general patterns predicted by memory-based theories,
607 whereas verb-final languages tend to exhibit the general patterns predicted by expectation-based theories”
608 seems to be difficult to maintain (also see Husain et al. (2015), for locality effects in Hindi). One
609 implication of our findings from Persian is that locality and expectation effects observed across studies
610 seem to be highly conditional on the language and syntactic construction being considered—broad
611 cross-linguistic generalizations may be difficult to make.

⁷ See Linzen and Jaeger (2015) for a recent empirical investigation of entropy in sentence comprehension using corpus data instead of sentence completion data. Linzen and Jaeger calculated entropy in several ways, and also evaluated another metric called entropy reduction (ER); however, we cannot evaluate ER here because that would require knowing the entropy for the word preceding the verb.

DISCLOSURE/CONFLICT-OF-INTEREST STATEMENT

612 The authors declare that the research was conducted in the absence of any commercial or financial
613 relationships that could be construed as a potential conflict of interest.

SUPPLEMENTARY MATERIALS

614 All items, data, and R code associated with this paper are available from <http://www.ling.uni-potsdam.de/~vasishth/code/SafaviEtAl2016DataCode.zip>.

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In review