

Journal of Experimental Psychology: Learning, Memory, and Cognition

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Online First Publication, July 5, 2018. <http://dx.doi.org/10.1037/xlm0000592>

CITATION

Rose, S. B., Aristei, S., Melinger, A., & Abdel Rahman, R. (2018, July 5). The Closer They Are, the More They Interfere: Semantic Similarity of Word Distractors Increases Competition in Language Production. *Journal of Experimental Psychology: Learning, Memory, and Cognition*. Advance online publication. <http://dx.doi.org/10.1037/xlm0000592>

The Closer They Are, the More They Interfere: Semantic Similarity of Word Distractors Increases Competition in Language Production

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Heterogeneous effects of semantic distance in language production have sparked a debate on the central assumption of many language production models, namely that lexical selection is a competitive process. In the present ERP study, we manipulated semantic distance in the picture word interference (PWI) paradigm systematically within taxonomic hierarchies. Target–distractor pairs were either closely related members of the same basic level category, hence sharing many semantic features (e.g., orangutan and gorilla), or distantly related members of the same superordinate category, sharing fewer features (e.g., orangutan and horse). Across related conditions, broad category membership (e.g., animals) was kept constant. Naming times reflected a systematic increase of semantic interference as semantic distance decreased. Early and later ERP modulations related to the semantic distance manipulation were observed at posterior regions starting at 234 ms and with an additional fronto-central cluster starting at 346 ms. Early effects are interpreted as indexing lexical selection while the late effects may reflect an N400-like component. Taking the behavioral and ERP modulations together, these results are in line with models of lexical selection that include an early competitive lexical selection process.

Keywords: language production, semantic distance, semantic interference, picture word interference, electrophysiology

Supplemental materials: <http://dx.doi.org/10.1037/xlm0000592.supp>

Inhibitory semantic context effects in picture naming tasks have long been taken as evidence for the competitive nature of lexical selection in language production. For instance, naming a picture (e.g., dog) is delayed in the presence of a semantically related distractor word (e.g., cat) relative to an unrelated word (e.g., pen; e.g., Damian & Bowers, 2003; Glaser & Döngelhoff, 1984; Glaser & Glaser, 1989; Hantsch, Jescheniak, & Schriefers, 2005; Schriefers, Meyer, & Levelt, 1990). According to lexical competition models (cf. Bloem & La Heij, 2003; Bloem, van den Boogaard, &

La Heij, 2004; La Heij, Kuipers, & Starreveld, 2006; Levelt, 1992; Levelt, Roelofs, & Meyer, 1999; Roelofs, 2018), target picture processing includes the coactivation of semantically related concepts and their corresponding lexical representations, which compete with the target for selection. Related distractor words contribute to this lexical competition by further enhancing the activation levels of nontarget lexical representations, resulting in delayed selection relative to unrelated words (e.g., Damian & Bowers, 2009; Dell'Acqua et al., 2010; La Heij, Heikoop, Akerboom, & Bloem, 2003; Melinger & Abdel Rahman, 2013; Roelofs, 1992; Schriefers et al., 1990; Vieth, McMahan, & de Zubicaray, 2014a).

One core assumption of lexical competition models is that competitive activation depends on the degree of semantic similarity (or distance) between representations. This factor determines the amount of activation spread within conceptual and between conceptual and lexical stages, where strongly related coactivated representations should compete more than weakly related representations due to high feature overlap (Dell, 1986; Roelofs, 1992; Vigliocco, Vinson, Damian, & Levelt, 2002; Vigliocco, Vinson, Lewis, & Garrett, 2004). Indeed, Vigliocco and colleagues (2004) reported modulated interference as a function of semantic distance in the PWI paradigm, with close distractors inducing stronger interference than more distantly related words. While similar semantic distance effects have been reported in subsequent studies using different naming paradigms (Aristei & Abdel Rahman, 2013;

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This research was supported by a grant (AB 277/4) from the German Research Council to Rasha Abdel Rahman. We thank Anna Eiserbeck, Johannes Rost, and Alina Karafiat for help with data acquisition and analysis.

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Navarrete, Del Prato, & Mahon, 2012, Experiment 3a and b; Vieth et al., 2014a, Experiment 2), others reported either no effects of semantic distance (Hutson & Damian, 2014; Navarrete et al., 2012, Experiment 2; Vieth et al., 2014a, Experiment 1) or faster naming in the context of close relative to distantly related distractors (Mahon, Costa, Peterson, Vargas, & Caramazza, 2007). These latter results challenge competitive models of lexical selection.

As an alternative to lexical competition models, Mahon and colleagues (2007) have formulated the response exclusion hypothesis to explain semantic context effects in the PWI task, including the facilitatory semantic distance effects they reported. Here, semantic relations may induce graded facilitatory, instead of inhibitory, effects. Specifically, according to this model, all semantic contexts induce facilitation due to semantic priming, with close semantic relations inducing more facilitation than distant semantic contexts due to higher semantic feature overlap with the target. Interference is assumed to originate at the postlexical stage of the articulatory output buffer, to which word distractors have privileged access. The output buffer constitutes a bottleneck that can be engaged with only one representation at a time; therefore, the distractor must be removed before the target word can be produced (see also Dhooge & Hartsuiker, 2012). The speed of exclusion depends on the response relevance of the distractor as a binary factor: Unrelated words are quickly dismissed because they are not eligible as response candidates. However, words sharing the broad semantic category with the target are potentially response relevant and therefore harder to reject. As a result, they are removed more slowly. Since the response relevance criterion is determined by the broad category membership, and is insensitive to graded differences in semantic distance (Mahon et al., 2007), close and distant distractors are equally response relevant; thus, exclusion times should be identical. Since close distractors induce stronger priming than distant distractors but take equal time to clear from the response buffer, they should be named faster than distant distractors, as was observed in Mahon and colleagues' experiments.

Irrespective of these theoretical aspects, one factor that may contribute to the inconsistency of semantic distance effects across studies is that diverse measures of semantic distance have been employed. For instance, two stimuli may be classified as closely related based on isolated shared features (e.g., the color: strawberry and lobster), even though they do not share the same semantic category and associated features (e.g., Hutson & Damian, 2014, Experiment 1; Mahon et al., 2007; Vieth et al., 2014a, Experiment 2; Vigliocco et al., 2002). The same stimuli may be classified as distantly related based on semantic similarity ratings that tend to underestimate distinctive features (Cree & McRae, 2003; Hutson & Damian, 2014, Experiment 2; Mahon et al., 2007; Vieth et al., 2014a, Experiment 1). Because shared and distinctive features may induce opposite effects in naming tasks (Vieth, McMahan, & de Zubicaray, 2014b; Rabovsky, Schad, & Abdel Rahman, 2016), measures that favor one over the other may lead to stimulus sets that induce effects of different polarity. Therefore, the types of semantic relation within and across categories may vary greatly between items within single studies as well as between different studies, and the specific measures used to define semantic relations may determine the polarity of the semantic distance effects.

In the present study, to create a more consistent stimulus set, we manipulated semantic distance systematically within taxonomic hierarchies (cf. Aristei & Abdel Rahman, 2013; Navarrete et al.,

2012; Rose & Abdel Rahman, 2016), thereby avoiding the aforementioned problems associated with heterogeneous materials. While superordinate category membership was held constant in the related conditions, the number of shared semantic features was manipulated as a function of the taxonomic relation between target and distractor. Specifically, in the distant condition, target and distractor are members of the same superordinate category but stem from different basic level categories, sharing few semantic features. In contrast, in the close condition, target and distractor are drawn from a common basic level category, sharing many features. For example, the orangutan shares only a limited number of features with other animals (e.g., horse), but many features with other members of the ape category (e.g., gorilla).

According to traditional competitive models of lexical selection (e.g., Levelt et al., 1999; see also Vigliocco et al., 2004), graded effects should be observed with stronger semantic interference for closely related relative to distantly related distractors, and fastest naming times for unrelated distractors. In contrast, according to the response exclusion hypothesis, close and distant distractors are members of the same superordinate category and are therefore equally response relevant; however, at the lexical level, close distractors should induce stronger semantic priming. Therefore, naming times should be faster in the close relative to the distant condition, and due to reduced priming, semantic interference effects are likely to be larger in the distant than the close condition, relative to the unrelated condition.

We used event-related potentials (ERPs) to gain further insight into the time course of semantic distance effects during naming. Competitive models assume that semantic interference should arise during early stages of word production (lexical selection), whereas the response exclusion hypothesis suggests that semantic interference should arise much later, close to articulation. Accordingly, lexical competition models predict early ERP modulations during lexical-semantic processing stages around 150 and 250 ms (Indefrey, 2011), while the response exclusion hypothesis predicts no such early modulations.

Empirically, semantic context effects in language production have been associated with ERP effects at anterior and posterior sites, in line with neuroimaging, stimulation, and patient studies reporting an involvement of (left) temporal and frontal regions (e.g., de Zubicaray, Wilson, McMahan, & Muthiah, 2001; Schnur, Schwartz, Brecher, & Hodgson, 2006; Schuhmann, Schiller, Goebel, & Sack, 2012; Wirth et al., 2011).

ERP studies investigating naming in a variety of semantic context paradigms have reported early context effects starting between 150 and 250 ms at posterior sites (e.g., Aristei, Melinger, & Abdel Rahman, 2011; Costa, Strijkers, Martin, & Thierry, 2009; Dell'Acqua et al., 2010; Maess, Friederici, Damian, Meyer, & Levelt, 2002; see also Indefrey & Levelt, 2004; Strijkers, Costa, & Thierry, 2010). Specifically, negative going ERPs corresponding to semantic effects have been reported in the cyclic blocking paradigm (homogeneous minus heterogeneous), associated with lexical-semantic processing (e.g., Aristei et al., 2011); likewise, a positive component around 200 ms has been related to lexical selection in the continuous naming task (later repetitions more positive compared to earlier repetitions; Costa et al., 2009; see also Rose & Abdel Rahman, 2017 for a similar positivity). Furthermore, examining the distractor frequency effect, Riès, Fraser, McMahan, and de Zubicaray (2015) observed early (100–300 ms)

ERP modulations at frontal and temporal scalp regions (high frequency more negative than low frequency), which they attribute to competition during lexical selection. Additionally, left frontal ERP modulations (more negative amplitudes for semantic relations/blocked conditions compared to unrelated/heterogeneous conditions) have been taken as evidence for enhanced a cognitive control mechanism during the resolution of lexical competition (e.g., Aristei et al., 2011).

In the PWI task, a relatively early positivity associated with related compared to unrelated masked distractors has been interpreted as an early onset of the N400 component and hence to reflect semantic priming, rather than lexical selection (Blackford, Holcomb, Grainger, & Kuperberg, 2012; Janssen, Hernandez-Cabrera, van der Meij, & Barber, 2015; see also Roelofs, Piai, Garrido Rodriguez, & Chwilla, 2016 for cross-lingual N400 effects in the PWI task). Indeed, several PWI studies have reported modulations of the N400 family with a fronto-central maximum between 300 and 500 ms and a more positive amplitude in the related relative to the unrelated condition (Blackford et al., 2012; Greenham, Stelmack, & Campbell, 2000). Since both lexical competition models and response exclusion models of semantic interference predict early semantic facilitation, these early N400-like modulations do not arbitrate between views. Instead, both approaches predict a second, additional effect. Early locus models predict another early modulation that indexes resolution of lexical competition (cf. Abdel Rahman & Melinger, 2009b), while late locus models predict a late modulation close to articulation. Critically, then, only early locus models predict modulations linked to semantic contexts that precede N400-like modulations.

Consistent with the former prediction, Piai and colleagues (Piai, Roelofs, Jensen, Schoffelen, & Bonnefond, 2014) have reported MEG evidence for a temporal overlap between priming and competition in the PWI paradigm. In line with the idea that conceptual and lexical processes largely overlap in time, they found interference to correspond to reduced activity between 350 and 650 ms at left-temporal regions, most likely reflecting semantic-lexical priming. Simultaneously, activity at left prefrontal regions increased for related versus unrelated distractors, analogous to the behavioral interference effect. Piai et al. suggested that this region was involved in the resolution of lexical competition. Hence, distinct ERP modulations reflecting lexical competition resolution and semantic priming may coexist with each being reflected in a different component, which contributes to the overall semantic context effects (see Abdel Rahman & Melinger, 2009a, 2009b, for a similar argument).

To summarize, according to lexical competition models, RTs should gradually increase from unrelated to distantly related to closely related words, whereas the opposite pattern should be found according to the response exclusion account. In ERPs, lexical competition models predict an early modulation around the time of lexical-semantic processing, which should be followed by an N400-like modulation. The response exclusion account predicts that the N400 priming effect should be the only early effect linked to semantic context, with semantic interference effects emerging closer to articulation. The study was approved by the local ethics committee.

Method

Participants

Twenty-four right-handed subjects, aged 19–30 years ($M = 23.4$, $SD = 3.7$) with normal or corrected-to-normal visual acuity and normal color vision participated for monetary compensation or for course credits. All participants were native German speakers and gave informed consent before experiment.

Materials

Stimuli were 125 color photographs of common objects scaled to 3.5 cm × 3.5 cm. The objects stemmed from five different superordinate categories (animals, clothes, tools, groceries, furniture), each subdivided into five basic level categories (e.g., animals: birds, fish, etc.; groceries: fruit, beverages, etc. see supplemental materials). The pictures were relatively easy to identify, and were typically not confused with other category members. The materials were selected to avoid strong visual similarities between members of small categories. We did so 1) by selecting pictures from different perspectives and avoiding unnecessary similarities and 2) by choosing close category members that are visually different (e.g., “eagle” vs. “owl”); please see supplementary materials for the complete list.

Each target object (e.g., eagle) was presented in three distractor conditions with a semantically close (owl), distant (gorilla), and unrelated (bed) word. In the close condition target and distractor were members of a basic level category (e.g., monkeys) sharing many semantic features; in the distant condition they were members of a superordinate category, sharing few features. All distractor words were part of the response set. The unrelated condition was constructed by rearranging related target–distractor pairs. The words were presented in red color and superimposed on the object pictures. Distractors were positioned so as not to obscure the image; hence, distractor placement differed between pictures. To avoid unwanted confounds, the distractor position for each picture remained constant across conditions.

Design and Procedure

Prior to the experiment, each participant was familiarized with the objects and their subordinate names. The objects' photographs were presented in random order on a monitor, and participants were asked to name each picture with its specific name at the subordinate level. If necessary, they were provided with the right name by the experimenter. After this procedure and while the EEG recording was prepared, participants were given sheets of paper with all pictures and their names printed below. Afterward, the PWI task started and participants were instructed to name the pictures on the monitor as fast and accurately as possible and to ignore the distractor words. Each trial began with a fixation cross displayed in the center of a light gray screen for 0.5 s. Then a picture–word pair was presented for 2 s ($SOA = 0$), followed by a blank screen for 1 s. Naming latencies were measured with a voice key during the entire duration of picture presentation. After the naming response was registered, the picture disappeared and the next trial followed. Each picture–word pair was presented two times, resulting in 750 trials, and a duration of about 40 min. The

presentation order of stimuli was randomized. After 40 trials, a pause screen was presented.

EEG Procedure

The continuous EEG was recorded with 62 Ag/AgCl electrodes arranged according to the extended 10/20 system, referenced to an electrode at the left mastoid. The sampling rate was 500 Hz. To register eye movements and blinks, electrodes were placed near the left and right canthi of the eyes and above and beneath the left eye. Electrode impedance was kept below 5 kOhm. Offline EEG was rereferenced using the average reference, and low-pass filtered (high cutoff = 30 Hz, 24 dB/oct). Eye movements and blink artifacts were removed employing the Multiple Source Eye Correction (MSEC) method implemented in BESA software (Berg & Scherg, 1994). In order to minimize possible distortion of the signal with cognitive sources, characteristic scalp topographies for blinks and eye movements were sampled for each participant individually during calibration trials (controlled eye movements in response to single tokens on the screen, e.g., a depicted left arrow to induce left eye movements; see Dimigen, Sommer, Hohlfeld, Jacobs, & Kliegl, 2011 for details and discussion). The resulting spatiotemporal patterns reflecting the artifacts were then subtracted from the raw EEG. Afterward, remaining artifacts were eliminated with an automatic artifact rejection procedure, excluding segments with potentials exceeding 50 μV voltage steps per sampling point and a threshold of 200 μV . The EEG data were

segmented in epochs of 2100 ms, starting 100 ms before the onset of the target. This 100 ms interval was used for baseline correction.

Because speaking can induce severe artifacts in the EEG (e.g., Brooker & Donald, 1980; Grözing, Kornhuber, & Kriebel, 1975; Wohlert, 1993), we employed a recently developed method for correcting the EEG signal from articulation-related artifacts (Ouyang et al., 2016; see De Vos et al., 2010; Porcaro, Medaglia, & Krott, 2015 for related approaches). Specifically, we used the residue iteration decomposition (RIDE) method that decomposes ERPs into separate component clusters with different trial-to-trial variabilities (e.g., stimulus-locked, response-locked, and latency-variable component clusters). Articulation artifacts can be separated from the EEG signal based on their large amplitudes and highly variable trial-to-trial latencies. In the study by Ouyang et al. (2016) the EEG and movements of the inner and outer vocal tract were coregistered using Electromagnetic Articulography (EMA). The authors showed that initially high correlations between articulator activity and speech artifacts in the EEG with a typical frontal-positive posterior-negative distribution dropped to almost zero after artifact correction, demonstrating successful artifact removal. The residue iteration algorithm (RIDE) used here decomposes ERP data into component clusters, which can be used to separate and reject articulation artifacts (Ouyang et al., 2016). We separated our data into a stimulus-locked S-component (which is equivalent to the corrected ERP and analyzed; search interval 0–500 ms after stimulus onset) and an R-component containing

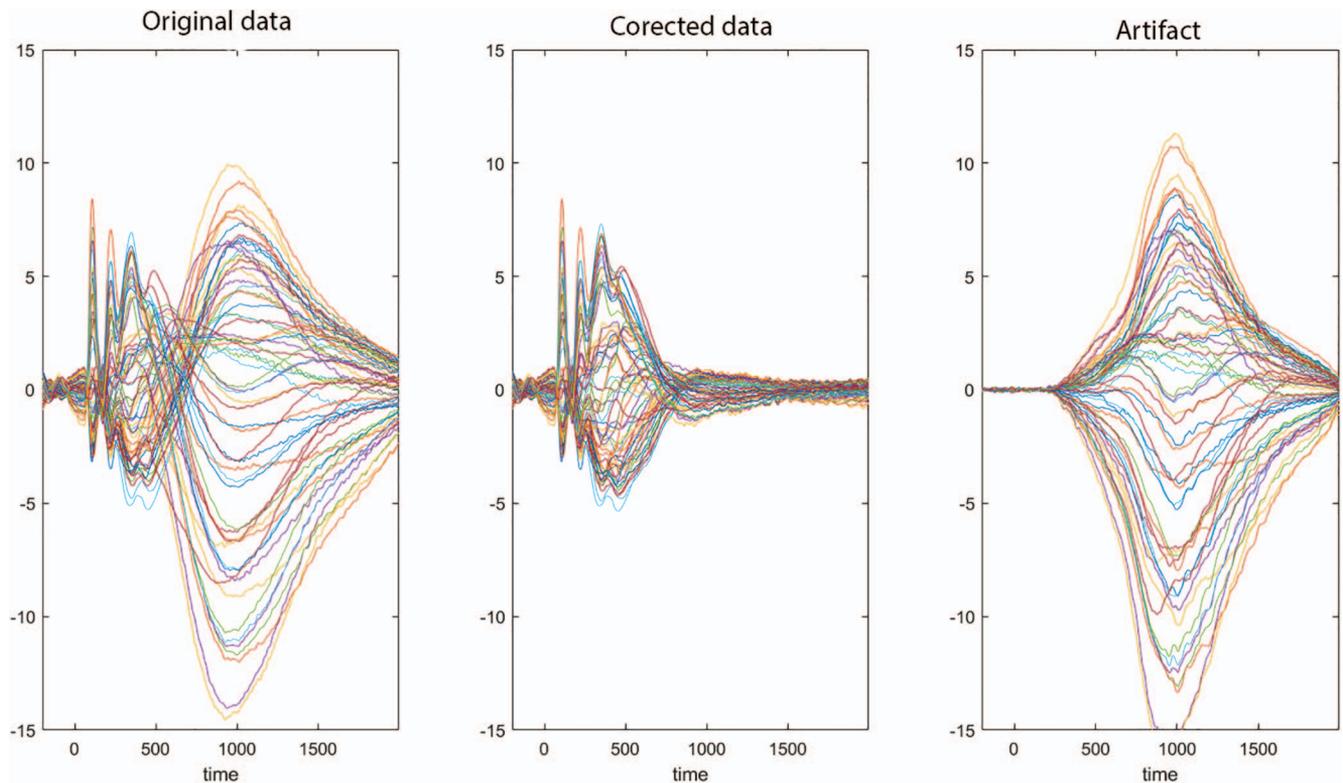


Figure 1. Left: Grand average ERP waveform before RIDE; middle: stimulus-locked component; right: response-locked component. See the online article for the color version of this figure.

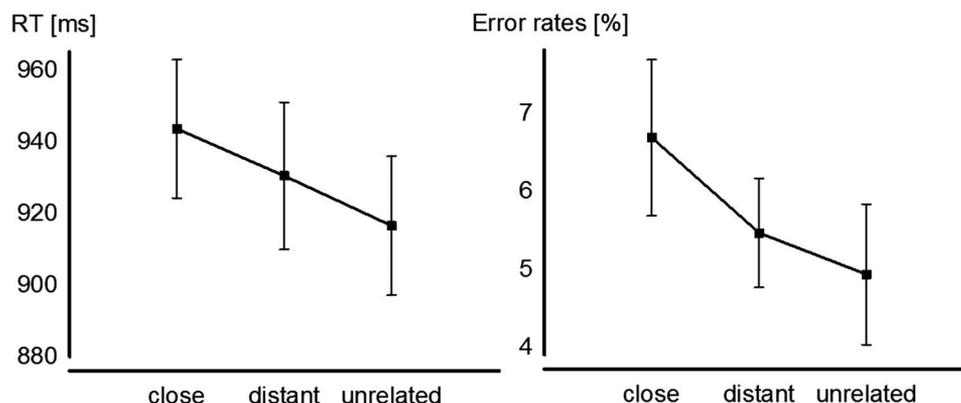


Figure 2. Mean naming latencies (left) and mean error rates (right) for each distractor condition. Error bars represent the standard error of the mean.

the artifact (which is rejected from the data; search interval 500–2500 ms after stimulus onset; see Figure 1).

Results

Behavioral Results

Naming times gradually increase from unrelated to distantly related to closely related distractor conditions (see Figure 2). A repeated measures analysis of variance (ANOVA) of naming latencies with the factor semantic distance (close, distant, unrelated), and participants (F_1) and items (F_2) as random variables, confirmed significant semantic distance effects, $F_1(2,46) = 21.9, p < .001, \eta_p^2 = .48$; $F_2(2,248) = 14.2, p < .001, \eta_p^2 = .1$. These were characterized by a linear trend, $F_1(1,23) = 38.7, p < .001, \eta_p^2 = .62$; $F_2(1,124) = 29.7, p < .001, \eta_p^2 = .19$, indicating that RTs increased linearly with decreasing semantic distance (close: $M = 943.53, SEM = 19.31$; distant: $M = 930.42, SEM = 20.49$; unrelated: $M = 916.50, SEM = 19.35$). Pairwise comparisons further confirmed this linear increase, showing significant differences between close and unrelated distractors ($t_1(23) = 6.2, p < .001$; $t_2(124) = 5.4, p < .001$), between distant and unrelated distractors ($t_1(23) = 3.5, p = .002$; $t_2(124) = 3.0, p = .003$), and between close and distant distractors ($t_1(23) = 3.2, p = .003$; $t_2(124) = 2.3, p = .023$).

A similar pattern was observed in the error rates, with increasing error rates as semantic distance decreases (close: $M = 6.68, SEM = 1.0$; distant: $M = 5.46, SEM = 0.7$; unrelated: $M = 4.93, SEM = 0.9$). Because of the categorical nature of the error data, errors were analyzed using mixed-effects logit models (cf. Jaeger, 2008) in R (Version 3.4.3, R Development Core Team, 2016) using the packages LanguageR (Baayen, 2013) and lme4 (Bates, Mächler, Bolker, & Walker, 2015). All participant responses, with correct trials coded as 0 and error trials coded as 1, were entered into a model with a fully specified random effect structure (Barr, Levy, Scheepers, & Tily, 2013). The fixed factor in this model was Semantic Distance (Close, Distant, Unrelated). The coefficient estimate (β), standard error (SE), Wald z -value (z) and p value were used to report the predictor parameters. The main effect of Semantic Distance was significant ($\beta = -0.20178, SE = 0.05021, z = 4.02, p < .001$). To assess the

incremental effect of semantic distance, pairwise analyses were also conducted. These results mirror the linear nature of the semantic distance effect observed in the reaction time (RT) analysis: Close versus Distant (MDiff = 1.22; $\beta = -0.11959, SE = 0.05788, z = 2.066, p = .038$); Distant versus Unrelated (MDiff = 0.53; $\beta = 0.13503, SE = 0.06868, z = 1.966, p = .049$); Close versus Unrelated (MDiff = 1.75; $\beta = -0.25629, SE = 0.06316, z = 4.058, p < .001$).

Electrophysiological Results

Only correct naming trials were included in the EEG analysis. Statistical analyses on error and artifact free data were performed with nonparametric cluster-based permutation tests (CBPT) as implemented in FieldTrip (Maris & Oostenveld, 2007, Version 20161024; with the function `ft_timelockstatistics`; for a recent application on which the present approach is based on, see Frömer, Maier, & Abdel Rahman, 2018) between 0 and 500 ms including each time point (2 ms) and all 62 electrodes, with 1000 randomizations using the FieldTrip MATLAB toolbox (Oostenveld, Fries, Maris, & Schoffelen, 2011) to determine time windows and electrode clusters that diverge between conditions.

The results are presented in Figures 3 and 4. The tests revealed significant differences between the close and unrelated conditions (Figure 3, top row, $p = .0009$ for the positive cluster, and $p = .02$ for the negative cluster; for details, see below), and significant differences between the close and distant conditions (Figure 3, middle row, $p = .003$ for the early positive cluster and $p = .02$ for the later positive cluster). However, there were no differences between the distant and unrelated conditions (Figure 3, bottom row, $ps > 0.45$). The effect of semantic distance can also be seen in Figure 4, with an augmented positivity in the close condition arising between 200 and 300 ms and between 400 and 500 ms.

Two clusters were identified to underlie the difference between the close and unrelated conditions. Starting at 234 ms and persisting until 480 ms, the closely related condition was associated with a stronger positive amplitude compared to the unrelated condition. This cluster emerges over posterior regions and then extends and broadens into central regions (Cluster 1; electrodes: Oz, O1/2, POz, PO3/4, PO7/8, Pz, P3/4, P5/6, P8, CPz, CP1/2, CP3/4,

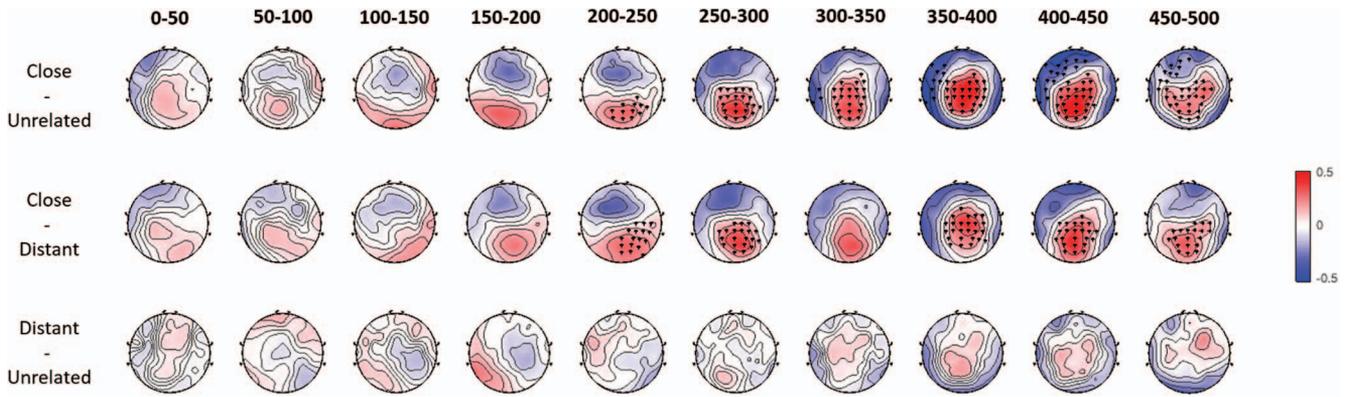


Figure 3. Topographies of the semantic distance effects. Maps show the difference between the semantically close versus unrelated (top), close versus distant (middle), and distant versus unrelated (bottom) conditions. See the online article for the color version of this figure.

CP5/6, Cz, C1/2, C3/4, C5/6, FC1/2, FC3/4, FC6, F4; $p < .001$). Starting from 346 ms and persisting for 112 ms, a second cluster emerges at left fronto-central regions (Cluster 2; electrodes: TP7, TP9, T7, FT7, FC5, F3, F5, F7, F9, AFz, AF3, AF7, Fpz, Fp1/2), with more negative amplitudes in the close relative to the unrelated conditions ($p = .02$).

The comparison between the close and distant condition revealed an early cluster, with a stronger positive amplitude associated with the close compared to the distant condition starting from 228 ms and extending until 292 ms at posterior electrode sites (Cluster 1; electrodes: Oz, O2, POz, PO3/4, PO8, PO10, Pz, P3/4, P6, P8, CPz, CP1/2, CP3/4, CP6, TP10, Cz, C2, C4, C6, T8; $p = .003$). After an

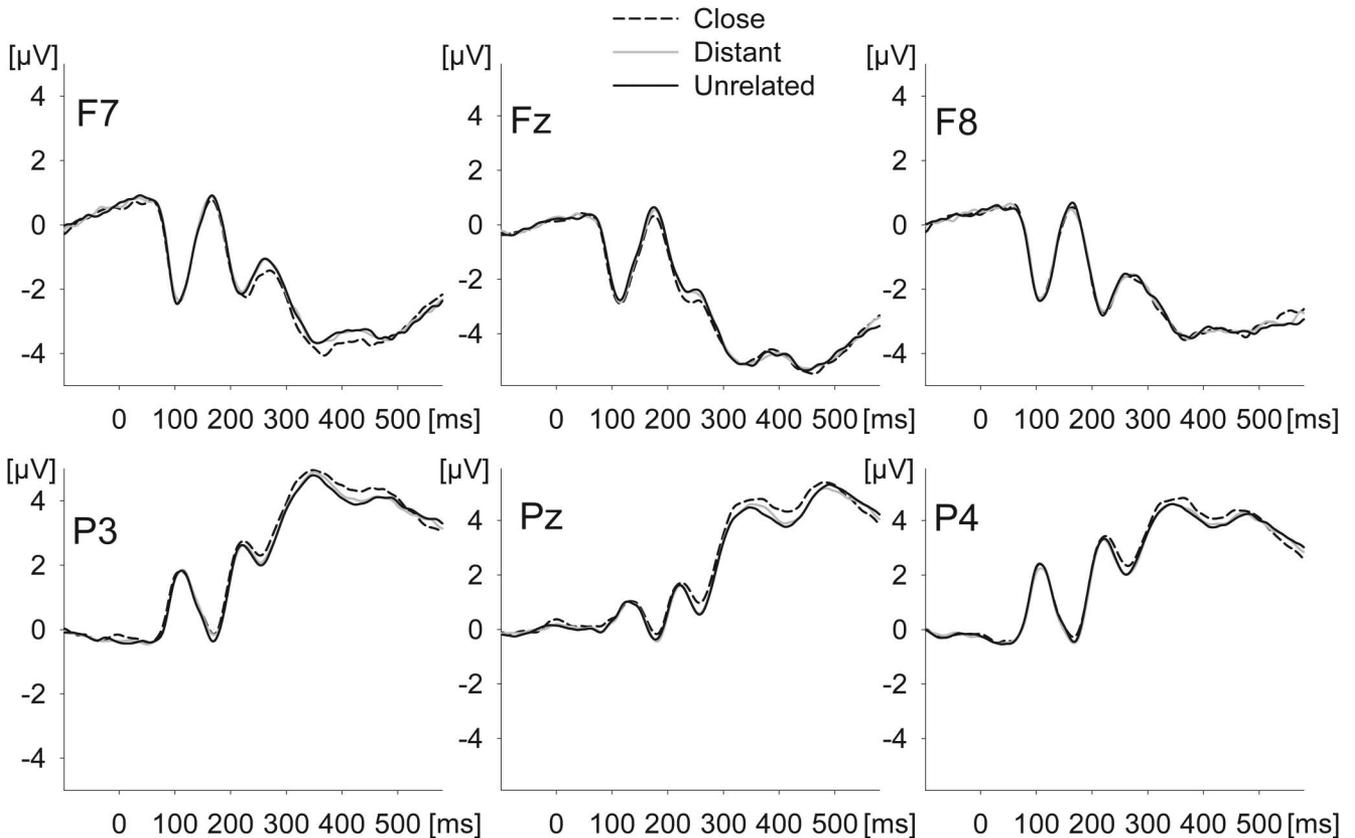


Figure 4. Effects of semantic distance on ERPs. Three electrodes from the frontal and posterior ROI are depicted.

interval of 54 ms in which no significant differences were observed, a second cluster, starting from 350 ms and extending for 124 ms, emerged with a centro-parietal distribution (Cluster 2; electrodes: Oz, O1/2, POz, PO3/4, Pz, P3/4, P5/6, P8, CPz, CP1/2, CP3/4, CP6, Cz, C1/2, C3/4, C6, Fc1/2, FC4, FC6, Fz; $p = .02$). Unlike the close versus unrelated comparison, no additional negative cluster at left fronto-central electrode sites was observed. The distant and unrelated conditions did not differ (cf. Figure 3).

Discussion

In this study, we explored semantic distance effects on object naming in a PWI task with behavioral measures and ERPs. We manipulated semantic distance systematically within taxonomic hierarchies, keeping broad category membership constant. In contrast to recently reported failures to find graded semantic distance effects (Hutson & Damian, 2014; Vieth et al., 2014a), and contrary to the observation of longer naming times in the context of distant compared to close distractors (Mahon et al., 2007), we found a gradual increase of semantic interference with decreasing levels of semantic distance. The slowest naming responses were associated with closely related distractors, intermediate responses with distantly related distractors, and fastest responses with unrelated words. This pattern is in line with some previous reports across different naming paradigms (Aristei & Abdel Rahman, 2013 in the PWI task; Vigliocco et al., 2004; see also Navarrete et al., 2012, Experiment 3; Rose & Abdel Rahman, 2017 in the continuous naming task; Vigliocco et al., 2002 in blocked cyclic naming).

This finding confirms predictions derived from traditional lexical competition models, assuming that the impact of semantically related words on lexical competition is augmented by the strength of activation spread between concepts as a function of their semantic feature overlap (Dell, 1986; Roelofs, 1992; Vigliocco et al., 2002, 2004). It cannot be explained by the noncompetitive response exclusion account (Mahon et al., 2007), which predicts the opposite pattern of interference in naming times, with slower naming in the context of distant compared to close distractors.

It also contrasts with an alternative attempt to account for the effects of Mahon and colleagues within a competitive framework. Abdel Rahman and Melinger (2009a, 2009b) argued in their model that the polarity of semantic context effects depends on the outcome of a trade-off between conceptual facilitation due to semantic priming and interference due to lexical competition. According to this swinging lexical network (SLN) account, the dominance of interference is determined by the activation of a lexical cohort that consists of an assembly of simultaneously coactivated semantic-lexical representations, and, crucially, the size of semantic interference is directly related to the size of the active lexical cohort. Thus, not only the activation strength of (single) competitors, but also the number of active competitors is assumed to affect lexical selection times and semantic interference. Assuming that closely related target–distractor pairs (e.g., owl, eagle) coactivate a category consisting of fewer members (birds) than categories coactivated by more distant target–distractor pairs (e.g., animals: owl, tiger), smaller interference in the close condition could be expected.

However, the present findings show that the *strength* of lexical coactivation, even of a relatively small cohort, is a crucial factor. If increasing cohort size goes along with a decrease in the strength

of mutual coactivation—as is the case for loosely related members of broad categories—the effect is weaker than the competition induced by a cohort of highly active competitors that coactivate each other, even if this cohort is of smaller size. Thus, it is the combined net effect of cohort size and activation strength, rather than either of these effects alone, that can explain the slower naming associated with close compared to distant distractors observed here. Specifically, closely related items share many specific features, and their activation spread converges on a small assembly of strongly interrelated and coactivated lexical representations, thereby intensifying lexical competition. In contrast, distantly related items share more global features and induce a wide but relatively unspecific activation spread between many loosely connected concepts—without strong converging activation patterns. Consistent with this view, there is a growing body of evidence demonstrating the independent importance of both cohort size and activation strength (e.g., Rabovsky et al., 2016; Rose & Abdel Rahman, 2016). Thus, the present finding adds to this growing literature by illuminating the interaction between cohort size and activation strength, thereby clarifying the swinging lexical network account.

While the results can be integrated into the SLN account, it seems harder to reconcile them with the response-exclusion account proposed by Mahon and colleagues (2007). Specifically, a fundamental claim of the original response-exclusion proposal is that the exclusion mechanism is tuned to assess response relevance and semantic distance, or semantic feature overlap is orthogonal to response relevance. It is this fundamental claim that allows the model to account for observations of semantic facilitation when traditional competition models predict interference. However, the present results would require the response exclusion account to accept that semantic feature overlap contributes to response relevance. By doing so, the time needed to exclude a distractor could be modulated by the closeness of the semantic relationship between target and distractor. However, it is unclear whether this revised conceptualization of the exclusion mechanism would still have the explanatory breath of the original proposal. Furthermore, to directly explain the strong inhibitory effects of semantic similarity found here, proponents would need additional assumptions on which of the opposing effects, lexical-semantic facilitation or response exclusion, would be stronger at any given time.

The graded RT effects found in the present study might be a consequence of the specific distance manipulation used. Most studies that failed to report graded semantic distance effects selected their stimuli according to feature generation norms (Hutson & Damian, 2014, Experiment 1; Mahon et al., 2007; Vieth et al., 2014a, Experiment 2; Vigliocco et al., 2002) or semantic similarity ratings (Cree & McRae, 2003; Hutson & Damian, 2014, Experiment 2; Mahon et al., 2007; Vieth et al., 2014a, Experiment 1). By using these measures, stimuli might be classified as semantically close even though they only share specific features (e.g., the color or shape) and stem from different semantic categories. Such inconsistencies can be avoided by using a taxonomic operationalization of semantic distance; by using superordinate and basic level categories, the membership to a common broad category is held constant between conditions.

In ERPs (see Figure 3), when comparing the closely related condition to the unrelated condition, we identified an early posterior cluster with a positive amplitude modulation (close minus

unrelated), starting at 234 ms. A topographically similar early modulation was also observed when comparing the close versus distant conditions. This was followed by a second positive cluster with a centro-parietal distribution (distant vs. unrelated). Finally, in addition to these positive modulations, we also observed an additional later negative cluster of left central and frontal electrodes in the close versus unrelated comparison, which was absent in the close versus distant comparison.

The onset of the early posterior effects, around 236 ms, is in line with other EEG/MEG studies manipulating semantic contexts in different naming paradigms (e.g., Aristei et al., 2011; Costa et al., 2009; Dell'Acqua et al., 2010; Maess et al., 2002; see Indefrey, 2011 for a metaanalysis). A very similar posterior positivity at about 200 ms has been reported in other semantic context paradigms and taken to reflect lexical selection (e.g., Costa et al., 2009; Strijkers et al., 2010). Indeed, Rose and Abdel Rahman (2017) used the same semantic relations and materials as employed here in a continuous naming task and, comparing the close versus unrelated conditions, reported a relative early posterior positivity very similar to the one observed here. Thus, in line with previous interpretations and predictions derived from lexical competition models, we conclude that the relative posterior positivity found here also reflects competitive lexical selection, with stronger competition in the closely related condition than is observed in either the distant or unrelated conditions.

The later positive modulation observed in the close versus distant comparison, Cluster 2, has many of the hallmarks of the N400 family: The modulations emerge at 350 ms over a cluster with a typical centro-parietal distribution. The N400 is associated with semantic processing and is commonly reported in Stroop and PWI experiments (Liotti, Woldorff, Perez, & Mayberg, 2000; Piai, Roelofs, & van der Meij, 2012; Shitova, Roelofs, Schriefers, Bastiaansen, & Schoffelen, 2016; Wong et al., 2017). Given the closer semantic relationship implemented in the close compared to the distant conditions, it was predicted that we would observe an N400 effect. However, an N400 was also predicted for the close versus unrelated condition, but in that comparison it is difficult to isolate a later effect because the early modulations are longer lasting.

The shorter duration of the early ERP modulations in the close versus distant comparison is in line with other reports and theoretical predictions (Indefrey, 2011). We can only speculate about the reasons for the longer duration in the close versus unrelated comparison. However, it is highly unlikely that the entire extended modulation observed in the close versus unrelated comparison, which extends for over 200 ms, reflects only lexical selection, which has been suggested to have a typical duration of approximately 75 ms (Indefrey, 2011). Therefore, we suggest that, like in the close versus distant condition, the positive modulations observed in the close versus unrelated condition actually reflect (at least) two temporally overlapping cognitive processes: the early modulation reflecting competitive lexical selection and a later modulation that reflects semantic effects of the N400 family (see also Piai et al., 2014).

The N400 modulation may specifically reflect the larger number of semantic features and neighbors shared between picture and distractor in the closely related condition compared to the unrelated condition, where there are minimal links between picture and distractor. In contrast, in the distant condition, the semantic rela-

tionships established between the picture name and distractor word are weaker, resulting in weaker semantic processing than in the closely related condition.

A third possibility is that the early posterior positivity observed in the close versus unrelated comparison may not reflect mechanisms of lexical selection at all, but rather fast feed forward processes that stabilize at later stages in the form of an N400 (Blackford et al., 2012; Janssen et al., 2015). Evidence against such an interpretation of our early positivity comes from the activation pattern observed in the close versus distant conditions. There, as discussed above, we see the emergence of two distinct ERP modulations, which mirror the effects proposed for the close versus unrelated conditions. It would be unlikely to observe modulations reflecting lexical selection in the close versus distant comparison but then observe similar modulations in the close versus unrelated comparison deserving of a differing interpretation.

One might ask why semantic effects should appear after lexical selection. We argue that the N400 is not the earliest brain activity related to semantic processing, and that at early points in time conceptual and lexical processes are strongly overlapping both temporally and spatially (e.g., Abdel Rahman & Melinger, 2009a, 2009b; Piai, Roelofs, Rommers, & Maris, 2015). Indeed, many studies on word processing suggest that access to semantic information starts earlier than the average latency or peak of the N400 component (e.g., Rabovsky, Sommer, & Abdel Rahman, 2012). Indeed, Thorpe, Fize, and Marlot (1996) demonstrated that basic semantic information can be available within 150 ms of picture presentation. Furthermore, the temporal regions where lexical selection is assumed to take place are also assumed to be the regions where semantic information and object concepts are processed (Visser, Jefferies, Embleton, & Lambon Ralph., 2012). Thus, conceptual and lexical processes largely overlap, with long-lasting conceptual processes starting before and continuing during and even after lexical selection, producing an overlap in ERPs (see Abdel Rahman & Sommer, 2003; Abdel Rahman, van Turennout, & Levelt, 2003 for evidence of parallel conceptual and phonological activation; Indefrey, 2011).

Finally, in the comparison between close and unrelated conditions, we observed a negative amplitude modulation at left fronto-central sites, starting at 346 ms. This activation may reflect enhanced cognitive control mechanisms during the resolution of lexical competition (cf. Aristei et al., 2011; de Zubicaray & McMahon, 2009; Schnur et al., 2006, 2009). However, this conclusion must be very tentative because many of the studies that have demonstrated activation in these regions used fMRI, which has poorer temporal resolution but much finer spatial resolution, allowing them to pinpoint specific brain regions. EEG, in contrast, has good temporal resolution but weak spatial resolution. Therefore, it is difficult to interpret EEG components on the basis of imaging research. For now, we simply point out that the negative modulation (close minus unrelated) over this left-lateralized cluster is unique to the close versus unrelated condition; no sign of a similar left-lateralized anterior effect is observed in the close versus distant comparison.

To summarize, we interpret the early relative posterior positivity as reflecting increased lexical competition in the close condition compared to either distant or unrelated conditions. The later more central modulations are interpreted as N400 type effects related to

the differences in the number of semantic features that are coactivated. We assume that an N400 is also present in the close versus unrelated comparison, although the individual components are harder to identify due to overlapping modulations. We speculate that the left-lateralized anterior negative modulations may reflect enhanced cognitive control mechanisms during the resolution of lexical competition in the close versus unrelated condition that may be particularly strong when distractor words are presented.

Theoretically, competition models of lexical selection assume that early effects resulting from lexical competition should emerge in the time window of around 200–250 ms (Indefrey, 2011; Indefrey & Levelt, 2004). In contrast, noncompetitive models of lexical selection such as the Response Exclusion Hypothesis (Mahon et al., 2007) predict that semantic context effects should only emerge at later time windows; no early effects are predicted. Both theories predict an N400-like modulation. Our early posterior modulations associated with the semantically close condition compared to the semantically distant and semantically unrelated conditions are therefore consistent with the predictions of competition models but more difficult to interpret within the context of a noncompetitive model.

In contrast to RTs, we did not observe graded ERP modulations. Specifically, even though distantly related distractors induced behavioral interference, no corresponding ERP modulations were found when comparing the distantly related distractor condition to the unrelated condition. We can only speculate about possible reasons. ERP differences between close and unrelated distractors were already relatively small and may have been too small to result in measurable ERP effects in the distant condition. In general, the available EEG studies seem to suggest that ERP modulations in the PWI paradigm are less pronounced than in other semantic context paradigms (cf. Hirschfeld, Jansma, Bölte, & Zwitserlood, 2008; Piai et al., 2012), such as the cyclic and continuous naming paradigms (e.g., Aristei et al., 2011; Costa et al., 2009), and other measures and techniques may be more suitable to detect distractor-mediated brain responses (e.g., Piai et al., 2012, 2014).

Taken together, the present findings are in contrast to the response exclusion account predicting reversed behavioral effects with faster naming times for close relative to distant distractors and no early ERP modulations preceding the N400. The findings of gradually increasing behavioral interference associated with decreasing levels of semantic distance and the early onsets of ERP modulations starting at about 230 ms—and that can be distinguished from later modulations of the N400 family—support language production models that incorporate competitive lexical selection. They furthermore suggest that the impact of semantic activation of competing lexical representations is strongly influenced by semantic feature overlap that can be viewed as a major determinant for semantic interference.

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Received August 15, 2016

Revision received February 21, 2018

Accepted February 22, 2018 ■