Intraspeaker Priming across the New Zealand English Short Front Vowel Shift

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Abstract

A growing body of research in psycholinguistics, corpus linguistics, and sociolinguistics shows that we have a strong tendency to repeat linguistic material that we have recently produced, seen, or heard. The present paper investigates whether priming effects manifest in continuous phonetic variation the way it has been reported in phonological, morphological, and syntactic variation. We analyzed nearly 60,000 tokens of vowels involved in the New Zealand English short front vowel shift (SFVS), a change in progress in which TRAP/DRESS move in the opposite direction to KIT, from a topic-controlled corpus of monologues (166 speakers), to test for effects that are characteristic of priming phenomena: repetition, decay, and lexical boost. Our analysis found evidence for all three effects. Tokens that were relatively high and front tended to be followed by tokens that were also high and front; the repetition effect weakened with greater time between the prime and target; and the repetition effect was stronger if the prime and target belonged to (different tokens of) the same word. Contrary to our expectations, however, the cross-vowel effects suggest that the repetition effect responded *not* to the direction of vowel changes within the SFVS, but rather the peripherality of the tokens. We also found an interaction between priming behavior and gender, with stronger repetition effects among men than women. While these findings both indicate that priming manifests in continuous phonetic variation and provide further evidence that priming is among the factors providing structure to intraspeaker variation, they also challenge unitary accounts of priming phenomena.

Keywords

Phonetics, sociolinguistic variation, priming, New Zealand English

Introduction

Sixty years of research in variationist sociolinguistics has provided us with a wealth of data and a reasonably detailed understanding of how many social and linguistic factors shape phonetic and

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phonological variation at the group level (e.g., Labov, 1994, 2001). The canonical methodology for this type of research is to perform some form of regression analysis to examine how various linguistic and extra-linguistic constraints operate on a single linguistic variable. One assumption underlying this approach is that each instance of variation is unique, unrelated to the preceding instance of the same variable from the same speaker (Tamminga, 2016). This assumption is at odds with findings from the literature in the so-called "third wave" of variationist sociolinguistics, which describes how individual speakers agentively use phonetic variants to construct different styles (Eckert, 2012, 2018). The main thrust of this literature is to claim that the meaning of a particular variant cannot be properly interpreted in isolation of the landscape of other variants with which it co-occurs. However, the methodological implication for variationist sociolinguistics is that instances of variation are, indeed, related to previous instances of variation, at least within a specific speech style. There have been some recent attempts to draw on these two sets of literature in order to better understand how multiple variants correlate and co-occur across large groups of speakers (e.g., Guy, 2013). This literature is concerned with correlating overall patterns of usage of variants across speakers; it does not look at usage as it unfolds in real time. Finally, there is a small but intriguing body of literature in corpus linguistics and sociolinguistics that considers the role of recency or "priming" as a motivation for the clustering of variants in real time in natural conversation. This literature aligns with the long-established finding from experimental psycholinguistics that when we talk or write, we have a strong tendency to repeat linguistic material that we have recently produced, seen, or heard (Bock, 1986; Branigan et al., 2000). This work has largely shown that repetition effects that occur in natural speech in real time are strikingly similar to those found in experimental studies of priming because they show that we have a tendency to repeat similar linguistic variants in quick succession in naturally occurring speech (e.g., Clark, 2018; Gries, 2005; Szmrecsanyi, 2006; Tamminga, 2016). Again, however, these studies have focused on the individual variable in isolation and simply asked: does a speaker's previous realization of a linguistic variable influence their immediately following realization of that same variable?

These four strands of literature have each provided new information about the factors that constrain and promote phonetic and phonological variation and change. From variationist sociolinguistics we have learned how individual variables pattern and change at the community level, and from the others we have learned that individual variables might cluster together and influence each other as discourse unfolds (either as speakers engage in identity work constructing speech styles or as they are biased by their own recently produced utterances). When combined, the four strands of literature raise new questions. In this paper, we are interested in understanding the repetition or priming effect in more detail. How does this priming effect—which has been reported to operate at the level of the individual variable—manifest when the variable in question is part of a larger systematic change in progress occurring at the community level, such as a vowel chain shift? Do we see priming effects only within phonological categories, or across phonological categories too? Finally, how does priming interact with social factors that also variously motivate or constrain a chain shift in progress?

The rest of the paper proceeds as follows. In the following section, we provide a more thorough review of some existing work on priming in sociolinguistics and corpus linguistics that led to the research questions we propose here. We then describe the variable phenomenon under investigation in this work: the New Zealand English (NZE) short front vowel shift (SFVS). We describe the corpus we used and the steps we took to create a measurement of "shiftedness": an index of participation in the vowel shift for each instance of each vowel. We present the results of a linear mixed-effects regression model; we structure the presentation of these results around our main research

questions and so the discussion of our results is interwoven with their presentation. Finally, we present some general conclusions.

2 Priming or repetition effects in natural conversation

There is now a long line of research from experimental psycholinguistics showing that when we talk, we tend to reproduce the same linguistic structures that we have recently produced or heard. Within this highly controlled experimental paradigm, participants' responses to a target stimulus are often influenced by an immediately preceding prime (e.g., Bock, 1986; Bock & Loebell, 1990; Branigan, et al., 2000). The question of whether this phenomenon exists to constrain variation in more natural dialogue has also received some attention. Gries (2005) explores variation in the British component of the International Corpus of English (which contains both spoken and written British English), and Szmrecsanyi (2005, 2006) examines repetition across five syntactic constructions using data from four large spoken corpora of English. Both studies find results that are in many ways similar to those from previous experimental work; at least with respect to the grammatical variables investigated, there is a clear and significant relationship between a speaker's/ writer's previous use of a variant and the likelihood of the next instance of that variable being of the same variant.

A handful of studies also exist that consider the effects of priming or repetition as a motivating factor in variationist sociolinguistics, again with a focus on morphosyntactic variation. Crucially, these studies also incorporate social motivations on variation into the analysis too. For instance, in Weiner and Labov's (1983) examination of variation in the agentless passive construction in English, the "mechanistic tendency to preserve parallel structure" (p. 56) was found to constrain this variation much more than any of the other social or linguistic factors considered in the analysis. Similarly, Poplack's (1980) study of variation in plural marking in Puerto Rican Spanish among immigrants in the USA found that "the presence of a plural marker before the token favors retention of that token, whereas absence of a preceding marker favors deletion" (p. 63). Scherre and Naro (1991, 1992) found a similar tendency with subject–verb agreement and subject–predicate adjective agreement in Brazilian Portuguese. Finally, Travis (2007) reports evidence of syntactic priming in sociolinguistic variation, this time in an analysis of variation in overt subject expression in two dialects of Spanish (New Mexican and Colombian).

Morphosyntactic priming studies have added to our understanding of the cognitive processes operating on language use, showing that where morphosyntactic variation exists in language, an alternative form, once used, has a greater chance of reuse next time. The extent to which these same processes operate to constrain phonetic and phonological variation in natural speech is less clear, largely because most previous work exploring this question (using a corpus-based methodology) has drawn on data from corpora of dyadic conversations, meaning this work did not control for the well-established accommodation effect wherein the speech of two people in conversation becomes more similar over time (Babel, 2010, 2012; Bell, 1984, 2001; Delvaux & Soquet, 2007; Giles & Coupland, 1991; Giles et al., 1991). For instance, Clark (2014), in a study of recency effects in TH-fronting, simply ignored all tokens produced by the interviewer in the analysis. Tamminga (2016, p. 343), in a study of variation in ING and T/D deletion, states that "accommodation is an important phenomenon in its own right," but that "its relationship to cross-speaker persistence is beyond the scope of this study." As a result, Tamminga "exclude[s] pairs of tokens that are interrupted by speech from other conversational participants." Neither of these approaches is terribly satisfying, as both fail to consider the possible (indeed, likely) effects of variants uttered by the interviewer or by an interlocutor on the following phonological variable, which may last beyond the next utterance.

There is one study that does attempt to disentangle potential priming effects from accommodation effects in a corpus of natural speech. Clark (2018) explored recency effects in phonology using speech data from a corpus of spoken monologues called the QuakeBox corpus (Clark et al., 2016). The QuakeBox project came about in the wake of the earthquakes that destroyed much of the city of Christchurch in New Zealand in 2010 and 2011. This collection of spoken stories is uniquely suited to this research question because members of the public were simply asked "tell us your earthquake story" and were then left alone to speak in a mobile recording studio. This means that we can reasonably establish the extent of within-speaker repetition without needing to consider any potential "interference" or accommodation towards an interviewer or interlocutor.

Despite the existence of previous work in corpus linguistics and sociolinguistics showing clear and replicable recency or repetition effects in natural speech, there has been hesitancy about labelling this phenomenon "priming." For instance, Szmrecsanyi (2005, p. 144) explains "I avoid using psycholinguistic terminology ('priming', 'prime', 'target', etc.) a priori because a corpus-based study may be inappropriate to explicitly investigate psycholinguistic mechanisms such as production priming effects." Similarly, Tamminga et al. (2016) and Tamminga (2019) avoid describing repetition effects in corpus data as "priming," as they argue that in corpus data it is non-trivial to disentangle repetition effects that are motivated by priming from those that are motivated by mechanisms like style-shifting, especially if repetitiveness is characteristic of the style being displayed by the speaker. It may also be the case that psychological factors, such as memory retrieval or attention, affect a speaker differently at different parts of a conversation, potentially influencing repetitive behavior in some parts of the discourse only. For this reason, Tamminga prefers to separate the notion of "priming," an underlying psycholinguistic explanation, from "sequential dependence," a mechanism in which "the outcome of a sociolinguistic alternation in one moment directly influences the likelihood of a matching outcome some moments later" (Tamminga et al., 2016, p. 33). In contrasting priming from sequential dependence, Tamminga rightly observes that repetition alone is not evidence for priming in corpus data.

Clark (2018) addressed this challenge by outlining two key signatures of priming that are well attested in the experimental literature, then looking for those signatures as interaction effects in a statistical model predicting the likelihood of repetition of a variable form in corpus data. The first signature of priming is the *decay* effect: repetition effects motivated by priming decay over time so that larger time depths between the prime and target should lead to a decrease in the repetition behavior (Goldinger et al., 1992). Clark (2018) showed that as the time interval increased between two instances of medial /t/ in a speaker's monologue, the likelihood of repetition between two instances diminished. The second signature of priming is the *lexical boost* effect. Clark (2018) showed that when two instances of medial /t/ occur sequentially, and both of those instances share the same word, if the first instance is realized as a voiced variant, the second instance is also more likely to be realized as voiced.[1] Clark (2018) explains that this interaction of lexical identity and phonological persistence is reminiscent of the lexical boost effect found elsewhere in research on syntactic priming (e.g., Hartsuiker et al., 2008; Jaeger & Snider, 2013). In this paper, we use the term *lexical boost* to refer to a possible strong priming effect when the prime and target share the same word (a usage that echoes Clark, 2018, and Tamminga, 2014, 2016).[2]

Based on previous findings of repetition effects in corpus data, we have reason to hypothesize that two consecutive instances of language in natural speech may indeed function as the prime and target in much the same way as they do in more controlled experimental studies. The remainder of this paper builds on the groundwork already laid by Clark (2018), and so we use the terms "prime" and "target" to refer to the immediately preceding instance of a vowel variable (the prime) and the current instance (the target), and we again look for the characteristic signatures of priming in our analysis.

We also note that in all of the work cited here, the approach has been to examine repetition or priming in the behavior of *a single variable*. Yet, we know that individual variables do not behave independently of each other, either because they cluster together to form speech styles and registers (e.g., Eckert, 2000) and/or because they are structurally related in an ongoing systematic change such as a chain shift (e.g., Labov et al., 1972). In this paper, we explore the evidence for priming as a motivating factor in the variation we see in the NZE SFVS.

3 The New Zealand English short front vowel shift

In NZE, the three short front vowel categories TRAP, DRESS, and KIT have changed their place of articulation fairly drastically since the 1890s, such that there are now no longer three short front vowels; rather, there are essentially two short front vowels with much higher and fronter nuclei than before (TRAP and DRESS) and one centralized vowel (KIT) (Langstrof, 2006).[3] The NZE SFVS is an example of what Labov calls a "regular sound change" or the "gradual transformation of a single phonetic feature of a phoneme in a continuous phonetic space" (Labov, 2010, p. 260). In other words, the phonological categories remain, but their realization changes gradually with no loss of lexical contrast. Furthermore, the changes in these three vowels are not independent of one another. Using evidence from archival recordings of speakers of NZE born in the late 1800s, Langstrof (2006) showed that the raising and fronting of TRAP began first, followed by the raising and fronting of DRESS as TRAP vowels encroached on the phonetic space of DRESS. The centralization of KIT is a later development and this is related to the encroachment of DRESS on the phonetic space previously occupied by KIT. The SFVS is therefore an example of a push chain. This complex set of interrelated changes is also variously constrained by several social and linguistic factors, such as a speaker's gender, age, and social class, and the phonological environment following the vowel and lexical frequency (Hay et al., 2015). Finally, it is important to point out that the SFVS is still a change in progress (Brand et al., 2019; Hay, et al., 2015), and potentially as a result, a fairly wide range of variability is still tolerated in NZE today. While overall mean F1 and F2 values for the SFVS vowels today are significantly different than they were 100 years ago, a synchronic NZE corpus will evidence considerable intraspeaker variation in TRAP, DRESS, and KIT.[4] This large range of intraspeaker variation, coupled with the fact that these three vowel variables are linked in a systematic change, makes the SFVS an interesting site to explore whether and how priming effects might constrain this type of sociophonetic variation. Following the methodology proposed by Clark (2018), we will seek to first establish the existence of repetition effects in the data, before exploring interactions between repetition effects and the two key signatures of priming: decay and lexical boost. Given that the SFVS plays out across three structurally linked vowel changes, we also consider whether and to what extent priming links variation in TRAP, DRESS, and KIT. Our research questions in this paper are as follows.

1. **Priming in the SFVS:** Is there evidence of intraspeaker priming in the NZE SFVS (taking the SFVS as a single unit)? For example, does a speaker's preceding realization of *any vowel* taking part in the NZE SFVS predict their next realization of *any vowel* in the SFVS? That is, does a preceding shifted vowel prime a following shifted vowel?
2. **Priming within and across vowel categories:** Is there evidence of intraspeaker priming both within and across vowel categories? For example, does a speaker's immediately preceding realization of a KIT vowel predict their next realization of a KIT vowel? Does a speaker's immediately preceding realization of a KIT vowel predict their next realization of DRESS? (Whereas RQ1 treats the SFVS as a single unit, RQ2 breaks this priming down into the individual vowel categories.) If there are priming effects here, are they stronger within or between vowel categories?

   3. **Priming and social factors:** Does intraspeaker priming interact with the social constraints
      that we already know to operate on the NZE SFVS?

We addressed these questions via a corpus study of spontaneous NZE speech. The following section details the data, dependent variable, and modeling procedure. We then reframe these research questions in terms of specific testable hypotheses.

4 Methods

4.1 Data and dependent variable

The data for this project came from the same source as Clark (2018): the QuakeBox corpus hosted by the New Zealand Institute of Language, Brain, and Behaviour at the University of Canterbury, New Zealand (Clark et al., 2016). We analyzed nearly 60,000 tokens of vowels involved in the NZE SFVS (KIT, DRESS, and TRAP) from a subset of the QuakeBox corpus consisting of 166 speakers' narratives (33 hours of speech).[5]

   Our approach to the data diverged from typical sociophonetic practice in two respects. Firstly, typical sociophonetic practice focuses on tokens that are likely to be "well-behaving" to the exclusion of other tokens; for example, Hay et al.'s (2015) study of NZE short front vowels included only stressed tokens from content words no longer than two syllables. In the context of priming, however, it makes sense to assume that *all* tokens are relevant for priming—not merely those that are convenient for researchers to measure. As a result, our data set included unstressed vowels and vowels in function words in addition to tokens that are more likely to be "well-behaving." We did exclude tokens that were *actually* "ill-behaving" (those that were misaligned by the forced alignment or mis-measured)—as opposed to those we merely suspect to be ill-behaving (due to being unstressed or in grammatical morphemes)—as these inaccurately measured tokens would distort our statistical analysis; we automatically excluded any tokens whose F1 or F2 measurements were more than two standard deviations outside the mean for that speaker and vowel category, and we hand-checked deviant tokens that remained after this exclusion to identify genuine outliers.

   Secondly, it is common practice in sociophonetic research to model vowel height and frontness separately (or to model one dimension to the exclusion of the other), typically via the first and second formants. In reality, F1 and F2 display considerable covariation; they do not act independently. This is particularly important to note in the context of a study on priming where we might expect that the previous production of a vowel might have an impact on the production of the next instance of a vowel in that same category; for example, we would predict that an immediately preceding instance of TRAP that is both higher and fronter than is typical for a given speaker will prime the following TRAP vowel to also be both higher *and* fronter. That is, we would not expect priming to only operate on either the F1 or F2 dimension alone. Furthermore, the NZE short front vowels have historically shifted along tilted axes in F1–F2 space, such that, for example, a lower KIT vowel is also likely to be more retracted (Gordon et al., 2004; Hay et al., 2015). As a result, we modeled vowel variation not in two separate statistical models of F1 and F2, but as a combination of F1 and F2 that we label the *shift index.* This is a single measure of a token's shiftedness with respect to the SFVS (i.e., along tilted axes in F1–F2 space). This approach, using a combination of F1 and F2, expands on the approach of Labov et al. (2013), who use the formula $-2*F1 + F2$ to model raising along the front diagonal of the vowel space; whereas Labov et al. apply the formula to multiple variables, we use vowel-specific coefficients for F1 and F2 derived from the community's patterns of variation.

Table 1. Measures used in calculating the shift index for each vowel category *v*.

	v		
	TRAP	DRESS	KIT
$\overline{SD_{F1,v}}$	87 Hz	74 Hz	106 Hz
$\overline{SD_{F2,v}}$	247 Hz	303 Hz	270 Hz
$\dfrac{\overline{SD_{F1,v}}}{\overline{SD_{F1,v}} + \overline{SD_{F2,v}}}$	−0.261	−0.197	0.282
$\dfrac{\overline{SD_{F2,v}}}{\overline{SD_{F1,v}} + \overline{SD_{F2,v}}}$	0.739	0.803	−0.718

The shift index was calculated as follows, schematized in Equation (1) and Table 1. We first extracted vowel measurements from the QuakeBox corpus then normalized F1 and F2 measurements using the Atlas of North American English method, with the default *G* value (Labov et al., 2006). Next, for each vowel *v*, we calculated each speaker's standard deviations for F1 and F2, and took the mean of those standard deviations ($\overline{SD_{F1,v}}$, $\overline{SD_{F2,v}}$). These means can be considered to be the average spread of tokens in F1 space and F2 space, so the sum of these two spread values can be considered the total average spread in the first two formants for vowel *v* ($\overline{SD_{F1,v}} + \overline{SD_{F2,v}}$). The coefficients for F1 and F2 were then calculated as the proportion of the total spread that each dimension accounted for ($\dfrac{\overline{SD_{F1,v}}}{SD_{F1,v} + SD_{F2,v}}$ for F1, $\dfrac{\overline{SD_{F2,v}}}{SD_{F1,v} + SD_{F2,v}}$ for F2). We assigned a negative sign to the F1 coefficients for TRAP and DRESS and the F2 coefficient for KIT, and a positive sign to the other coefficients, to reflect the fact that higher/fronter TRAP/DRESS tokens and lower/backer KIT tokens are more advanced in the SFVS (and therefore should have greater shift index values). Finally, we standardized (*z*-scored) the result of this linear combination within each vowel in order to make model calculations more tractable and place the shift index distribution for each vowel on the same scale

$$ShiftIndex_{v,i} = standardize\left(\frac{\overline{SD_{F1,v}}}{SD_{F1,v} + SD_{F2,v}} F1_i - \frac{\overline{SD_{F2,v}}}{SD_{F1,v} + SD_{F2,v}} F2_i \right) \qquad (1)$$

As a result of this calculation, each token of each vowel had a single measure of shiftedness with respect to the SFVS. Calculating and standardizing the shift index separately for each vowel permitted apples-to-apples comparisons of the shiftedness of vowel realizations, even if they belonged to different vowel categories. Note that we calculated the shift index *between* speakers rather than within speakers, meaning that, for example, a KIT token with F1 and F2 values of 524 and 1588 Hz had a shift index of 0.5 regardless of speaker or speaker gender; the advantage of this between-speaker calculation is that we were able to model the effect of social factors (e.g., gender) on this variation, and so consider whether social factors interact with the repetition effects.

To help visualize how variation in F1–F2 space translated to shift index variation, Figure 1 displays shift index axes for each vowel category. The symbols TRAP, DRESS, and KIT represent the mean of all speakers' normalized F1 and F2 for those vowel categories; in other words, those are the points at which the shift index equals zero for that vowel category (since raw shift index values were *z*-scored
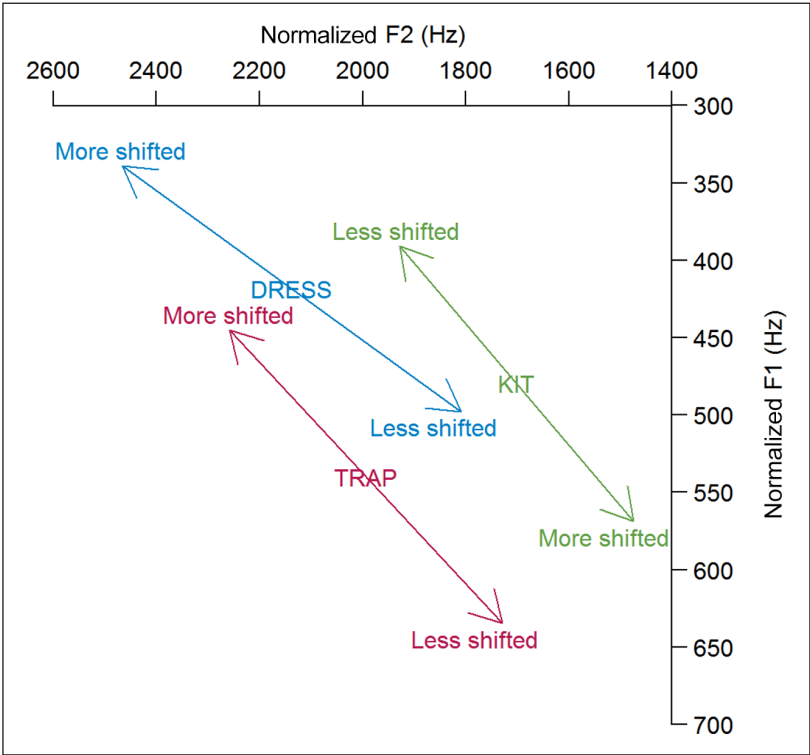
Figure 1. Shift index axes for the vowel categories in this study. For each vowel, the arrowhead at "Less shifted" marks a hypothetical token with a shift index of −1, the vowel symbol marks a token with a shift index of 0, and the arrowhead at "More shifted" marks a token with a shift index of +1; tokens varying perpendicularly to these axes would share the same shift index values. The KIT axis proceeds in the opposite direction to the TRAP/DRESS axes because KIT centralizes in New Zealand English, whereas TRAP/DRESS raise and front, and thus a more centralized KIT token is more shifted.

within each category). Within each vowel category, shift index values vary parallel to the shift index axis, and shift index values are invariant perpendicular to the axis. As a result, the slopes of the shift index axes indicate the relative contribution of F1 versus F2 to the shift index; TRAP and KIT have steeper shift index axes than DRESS, since TRAP and KIT have proportionally more spread in F1 than DRESS (Table 1). While shift index values overall ranged from −3.98 to +3.69, the arrows in Figure 1 end at the points for which the shift index is ±1 for that category (to give a sense of the magnitude of different shift index values); the arrowhead at "Less shifted" marks a hypothetical token with a shift index of −1, and the arrowhead at "More shifted" marks a token with a shift index of +1. Finally, the ends of the arrows are labelled to reflect the signs of the shift index axes; for example, since lower/backer KIT tokens are more advanced in the SFVS, the lower/backer arrowhead of the KIT shift index axis is labelled "More shifted" and corresponds to a shift index of +1.

After calculating the shift index, we made a number of additional exclusions to the data. Firstly, since we modeled repetition by looking at how each token was affected by the immediately preceding token, we excluded each speaker's first short front vowel token because the immediately preceding token, occurring before the recording started, was unknown. For the same reason, our model excluded outliers and tokens that immediately followed outliers. That is, if token i was an outlier, then the target shift index for token i was unreliable, and since the target shift index for token i
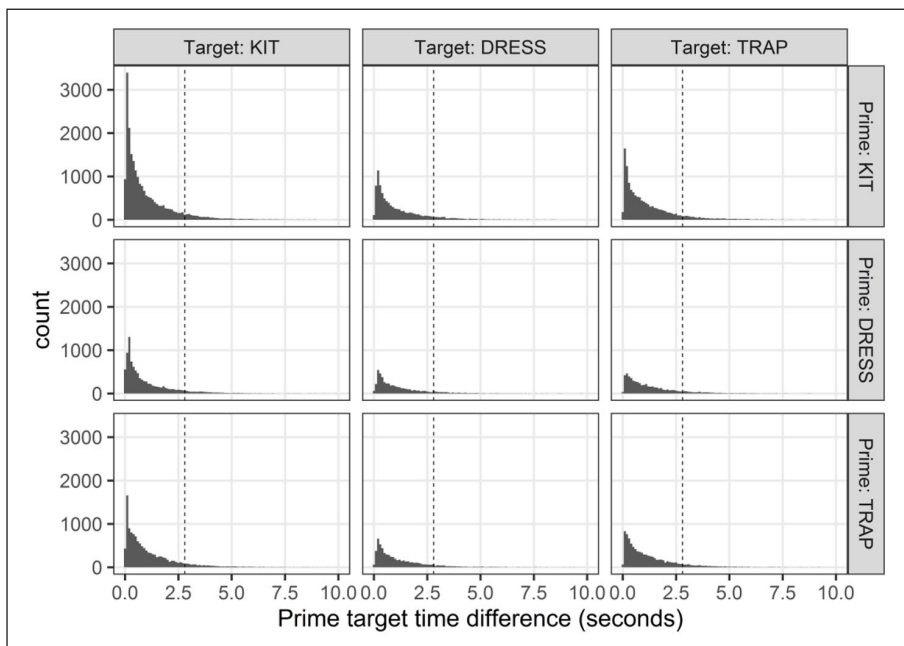
Figure 2. Prime target time difference distributions by target vowel category and prime vowel category. The dotted line marks the 90th percentile of overall distribution (2.8 seconds); tokens with a prime target time difference longer than this cutoff were excluded from the model.

served as the prime shift index for token $i + 1$, both tokens were excluded; since this did not affect the target shift index for token $i + 1$ (i.e., the prime shift index for token $i + 2$), token $i + 2$ was not excluded. Secondly, because we assume that exemplars are constructed at the level of the individual word (e.g., Docherty & Foulkes, 2014), we excluded tokens whose prime was in the same word as the target (e.g., the target TRAP vowel in *liquefaction* was excluded because the previous short front prime token, KIT, was in the same word). Thirdly, because one predictor in our model was the natural logarithm of the time difference between the prime and target, a handful of targets that immediately followed their prime were excluded because their prime target time difference was zero, and the log of zero is undefined. In addition, targets that were more than 2.8 seconds after their prime (the 90th percentile of prime target time difference—see Figure 2) were excluded, as these represented high-leverage points that had the potential to unduly influence the effect of prime target time difference. Fourthly, since our final model found the morphemic status of the prime (grammatical versus lexical) to be a significant predictor, we excluded targets whose prime's morphemic status was ambiguous in the absence of hand-coding the data (*had*, *have*, *has*, *back*). In all, from an original sample of 91,146 non-outlier tokens, we analyzed 59,824 tokens (Table 2).[6]

4.2 Modeling procedure

We carried out statistical analysis via mixed-effects modeling in R (R Core Team, 2018) using the package lmerTest (Kuznetsova et al., 2016), which calculates significance levels using Satterthwaite's (1946) approximation for degrees of freedom. The target shift index was the dependent variable in these models; for categorical predictors, contrasts were coded via treatment coding (i.e., treating one level as the "baseline" level). We started with a baseline model that was

Table 2. Distribution of total number of instances of prime–target combinations in the analysis data set.

		Prime			
		KIT	DRESS	TRAP	Total
Target	KIT	14,411	5902	6871	27,184
	DRESS	6437	3580	3704	13,721
	TRAP	9439	4151	5329	18,919
	Total	30,287	13,633	15,904	59,824

designed to address RQ1 and RQ2 (i.e., assessing priming in the SFVS, within and across vowel categories). This model included fixed-effects predictors corresponding to priming effects (RQ1): prime shift index (repetition effect), prime target time difference (decay effect), and prime target same word (lexical boost effect). The baseline model also included the vowel categories of the prime and target as fixed-effects predictors (RQ2), as well as random intercepts and random prime shift index slopes for individual speakers and target words. All of the fixed-effects predictors in the baseline model yielded significant effects.

We then tested additional predictors in two rounds. In order to address RQ3, we tested social predictors available in QuakeBox: gender and age group. We then tested internal predictors to control for additional phonetic, phonological, morphological, and lexical effects influencing variation in the SFVS. Predictors were only added if the resultant model converged, did not suffer from excessive collinearity,[7] and significantly improved upon the fit of the previous model (assessed via a likelihood ratio test at $\alpha = .05$); if multiple candidate predictors satisfied these criteria, the predictor that yielded the greatest improvement in model fit was added first, as measured by the Akaike information criterion (Akaike, 1974). When predictors were added to the model, by default they were added as four-way interactions with the prime shift index, target vowel category, and prime vowel category; where possible, these interactions were reduced to lower orders based on whether the higher-order terms significantly improved model fit. Table 3 shows the predictors that we tested by round. We report only the results from the final model.

We note that while collinearity (the dependence of one predictor on another) is to be avoided in any multiple regression model as a matter of statistical due diligence, it was especially important in the present study given our research questions about the phonetic shape of the target being influenced by that of the prime. Models that suffer from collinearity overstate the effects of the collinear predictors; by avoiding collinearity in our models, we eliminated the possibility that any apparent priming effect was merely a statistical artifact. This statistical due diligence is thus necessary because it preserved the model's inferences for the predictors corresponding to our core hypotheses (see below). We do recognize, however, that readers may wish to entertain different hypotheses about the data that spotlight different predictors than we did, given that we were unable to assess significance for predictors excluded due to collinearity. To that end, we have made the data for this analysis available at https://github.com/nzilbb/Priming-SFVS-Data so readers can assess alternative modeling procedures.

5 Hypotheses

Based on our research questions, we propose the following hypotheses.

1. **Priming of the SFVS:** We expect to observe overall effects consistent with signatures of priming: repetition, decay, and lexical boost. That is, we expect that the immediately

Table 3. Predictors tested as fixed effects; italicized names refer to different rounds of adding predictors to the model. For categorical predictors, the first level listed is the baseline level.

Predictor	Predictor type
Baseline model	
Prime shift index	Continuous (z-scored)
Prime target time difference	Continuous (z-scored)
Prime target same word	Categorical (TRUE, FALSE)
Target vowel category	Categorical (TRAP, DRESS, KIT)
Prime vowel category	Categorical (TRAP, DRESS, KIT)
Social predictors	
Age group	Categorical (36–45, 46–55, 56–65, 66+)
Gender	Categorical (female, male)
Control predictors	
Target vowel duration	Continuous (z-scored)
Prime vowel duration	Continuous (z-scored)
Target morphemic status	Categorical (grammatical, lexical)
Prime morphemic status	Categorical (grammatical, lexical)
Speech rate	Continuous
Target word corpus frequency	Continuous (z-scored)
Prime word corpus frequency	Continuous (z-scored)
Target word syllable count	Continuous
Prime word syllable count	Continuous
Prime target same phrase	Categorical (TRUE, FALSE)

preceding value of the prime shift index should be a significant predictor of the following target shift index (repetition), regardless of vowel category. There should also be an interaction of prime shift index and prime target time difference such that the greater the time difference between prime and target, the weaker the overall priming effect (decay). In addition, there should be an interaction of prime shift index and prime target same word such that the priming effect is stronger in cases where the prime and target vowels occur in a repeated lexical item (lexical boost). This trio of patterns (repetition, decay, and lexical boost) would replicate the results presented of Clark (2018), in which the modeling was performed on a categorical variable (voiced versus voiceless), by extending those findings to a continuous variable (vowel chain shift).

2. **Priming across and within vowel categories:** We expect that the intraspeaker priming effects should persist both within vowel categories (e.g., TRAP prime and TRAP target) and across categories (e.g., TRAP prime and KIT target). We also expect a particular direction for this cross-category priming: regardless of the vowel category of the prime and target, we predict that more-shifted tokens should prime more-shifted tokens (and less-shifted should prime less-shifted). Because the shift index increases in the direction of shift regardless of vowel category (Figure 1), this expectation should manifest as positive slopes in all cases of cross-category priming. Finally, we expect that within-category priming should be stronger than cross-category priming.

3. **Priming and social factors:** We expect to find some interaction between social factors and priming behavior. Many priming studies report individual participant differences in priming behavior (e.g., Andrews & Lo, 2013; Medeiros & Duñabeitia, 2016; Ortells et al., 2016), but the explanations given for these differences are cognitive, not social (e.g.,

differences in participants' working memory capacity or reading profile). However, Clark (2018) found that certain variants of medial /t/ were more subject to repetition effects among women and other variants were more subject to repetition effects among men. This was attributed to the fact that medial /t/ is a socially conditioned change in progress in NZE. Clark (2018) argued that some variants of medial /t/ have more social salience for women and other variants have more social salience among men. This heightened social salience (or "socially weighted encoding," in the framing of the model of speech perception proposed by Sumner et al., 2014) could cause these variants to have heightened levels of activation among different groups in the population, raising their chance of re-selection. The NZE SFVS is also a socially conditioned change in progress and so we expect that there may be some interaction between social predictors (age and gender) and priming predictors (recency, decay, and lexical boost) in our model.

As described in the preceding section, our modeling procedure was designed to test these hypotheses by first starting with the baseline predictors in Table 3 (corresponding to RQ1 and RQ2), then testing social predictors (RQ3), then testing control predictors that influence variation in the SFVS.

6 Results

Our model of the target shift index includes the baseline predictors (prime shift index, prime vowel category, target vowel category, prime target time difference, prime target same word), one social predictor (gender), and two control predictors (prime morphemic status, speech rate).[8] There were numerous significant higher-order interactions among these predictors, as the interaction of the prime shift index and target vowel category interacted with all other predictors. Appendix A details the final model structure and output. As we discuss in more detail below, the model broadly indicates support for hypotheses 1 and 3, and partial support for hypothesis 2.

Readers should note that, as a result of the criteria for adding predictors, the exclusion of a given effect from a model does not necessarily indicate nonsignificance; predictors were not assessed for significance if adding them to the model caused non-convergence or excessive multicollinearity, and in fact, all of the predictors in Table 3 that were not included in the model were excluded due to multicollinearity. In other words, it is possible that some effects that influence variation in these vowels in this community are nevertheless excluded from the model because their inclusion causes the model to be untenable. To that end, we have made the data for this analysis available at https://github.com/nzilbb/Priming-SFVS-Data so readers can test how the modeling pans out differently under assumptions other than those that guided our modeling procedure.

In our model of the target shift index, the prime shift index was significant in all two-way and three-way interactions with the prime vowel category and target vowel category ($ps < .001$). As a result of these interactions, the main effect of the prime shift index is not meaningful on its own, so in the following section we address RQ1 and RQ2 at the same time.

6.1 RQ1 and RQ2: Priming of the SFVS, across and within vowel categories

The model of the target shift index indicated evidence of effects—repetition, decay, and lexical boost—that are key signatures of priming, across the SFVS as a whole. In other words, our findings indicate evidence to support hypothesis 1, as we demonstrate in this section. With respect to RQ2, these priming effects were also present both across and within vowel categories. However, these results ran counter to other expectations in hypothesis 2. Most notably, the cross-vowel priming results for SFVS vowels changing in opposite directions (TRAP/DRESS raising/fronting versus KIT
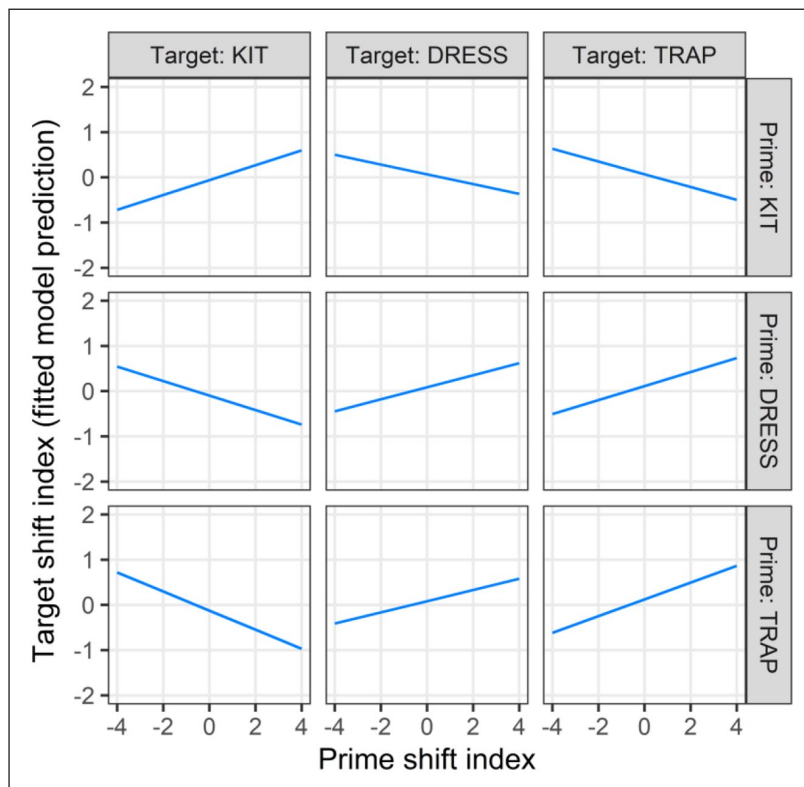
Figure 3. Predicted target shift index by the prime shift index, target vowel category, and prime vowel category.

centralizing) strongly suggest that the driving factor is *not* the direction of the SFVS, but rather peripherality.

In Figure 3, the effect of the prime shift index on the target shift index is broken down into the nine possible prime–target combinations in our data. If hypothesis 2 were correct (more-shifted tokens should prime more-shifted tokens), a positive relationship of the prime shift index and target shift index should be replicated in all nine facets, given that the shift index increases in the direction of shift regardless of vowel category (Figure 1). Instead, this positive relationship is replicated in just five facets: all three within-category prime–target combinations, plus two cross-category combinations (TRAP primes with DRESS targets, and DRESS primes with TRAP targets). For example, a more advanced (i.e., fronter and higher) TRAP primes a more advanced (i.e., fronter and higher) DRESS. However, the opposite pattern is found for KIT primes with TRAP/DRESS targets and TRAP/DRESS primes with KIT targets; for example, a *less* advanced (i.e., fronter and higher) KIT primes a *more* advanced (i.e., fronter and higher) TRAP. In other words, while a repetition effect does occur both within and across vowel categories, it appears to be sensitive not to how shifted the prime and target are, but how peripheral they are. Moreover, the magnitude of this repetition effect exhibits little variation between prime–target combinations, contrary to our expectation (in hypothesis 2) that within-category priming would be stronger than cross-category priming.

The effect of the prime shift index on the target shift index was mediated by two other priming predictors: prime target time difference (decay) and prime target same word (lexical boost). The prime
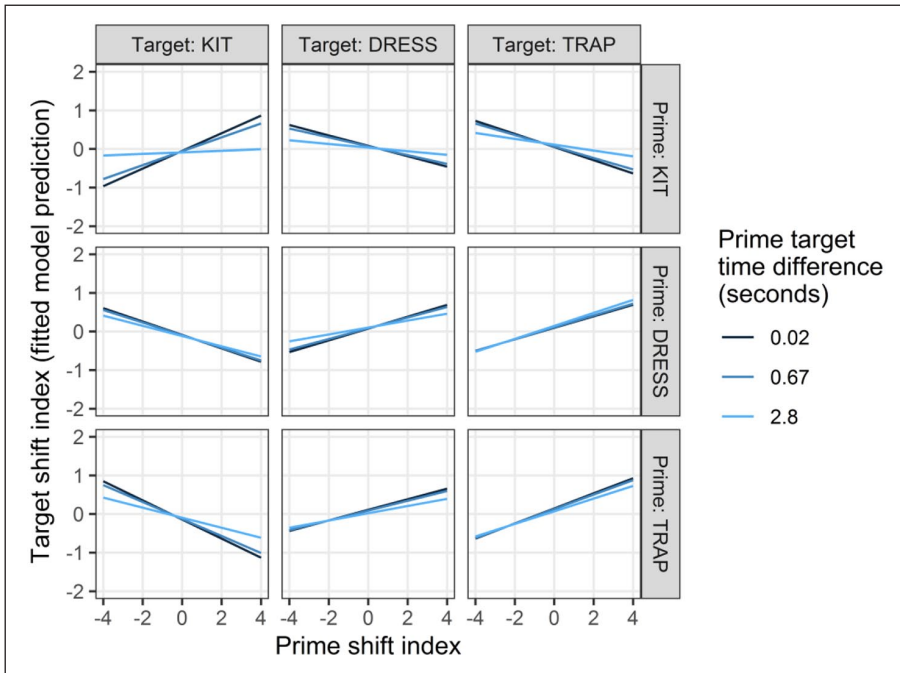
Figure 4. Predicted target shift index by the prime shift index, target vowel category, prime vowel category, and prime target time difference. The prime target time difference values at which lines are plotted represent the minimum, median, and maximum time differences in this data. (Color online only.)

shift index had a significant negative interaction with prime target time difference, indicating that the strength of the repetition effect displayed in Figure 3 is mitigated by the amount of time between the prime and target. In particular, we can interpret the slope of the line in the target shift index–prime shift index Cartesian space as indicating the strength of the repetition effect, with steeper lines indicating a stronger effect of the prime shift index on the target shift index and thus a stronger repetition effect. Given that visual interpretation, Figure 4 demonstrates the strongest repetition effect when the prime target time difference is at its minimum, 20 milliseconds (since the darkest line has the steepest slope), a weaker effect when the prime target time difference is at its median, 670 milliseconds (since the medium-blue line has a shallower slope), and a weaker effect still when 2.8 seconds separate the prime and target (since the lightest line has an even shallower slope, nearly leveling off completely in some cells). In other words, these data exhibit a decay effect characteristic of priming.

This interaction was significant across all three-way and four-way interactions with the prime vowel category and target vowel category ($ps < .001$), meaning that this decay effect was present across all prime–target combinations, albeit with stronger decay effects for some prime–target combinations and weaker decay effects for other combinations. Moreover, this decay effect proceeds rather rapidly. The repetition effect is strongest at a 20 millisecond prime target time difference; when the time difference grows to merely 670 milliseconds, the repetition effect diminishes considerably, as evidenced by the shallower slopes of these lines.

The prime shift index also had a significant positive interaction with prime target same word, with a stronger repetition effect when the target word was a repetition of the prime word. In other words, these data exhibit a lexical boost effect characteristic of priming. Unlike the repetition and decay effects, this lexical boost effect does not appear to be mediated by the vowel category of the prime or
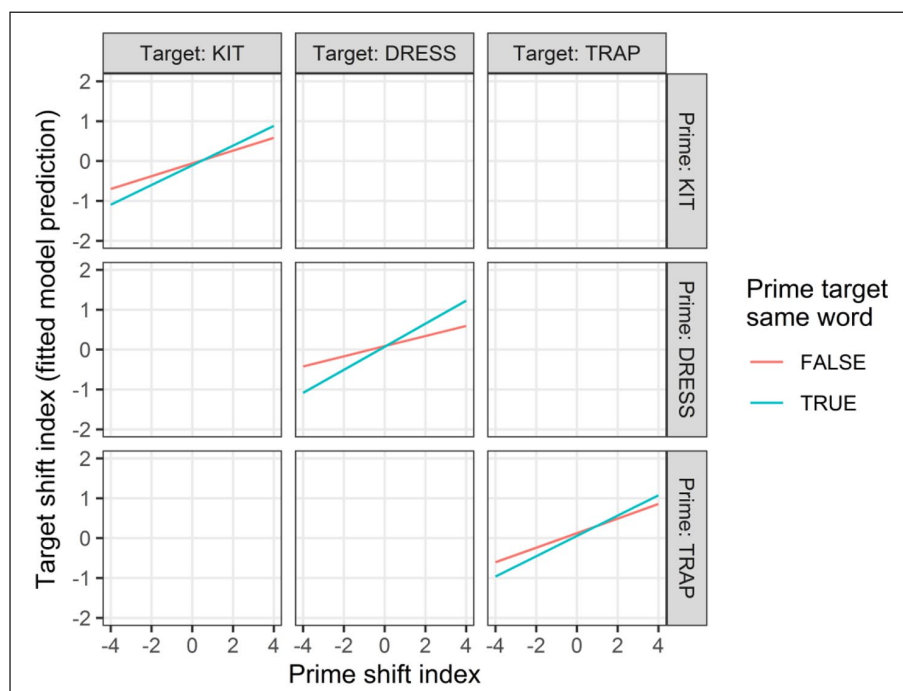
Figure 5. Predicted target shift index by the prime shift index, target vowel category, prime vowel category, and prime target same word.

target (although Figure 5 suggests a weak trend in this direction), as there were no significant three-way interactions between the prime shift index, target vowel category, and prime target same word ($ps > .12$).[9] Taken together, these results indicate support for hypothesis 1, as we find repetition, decay, and lexical boost effects that are often taken to be key signatures of priming in speech.

Readers may wonder whether the effect of the prime shift index is driven by *inter*speaker variation rather than *intra*speaker priming, with some speakers having more shifted vowels overall than other speakers and all speakers hitting their own targets repeatedly. Mitigating against this interpretation is our inclusion of speaker as a random effect in the model. If the variation in our data were merely interspeaker variation, then the random intercept of speaker would "soak up" all of this variation, leaving nothing for the effects of the prime shift index to account for (Tamminga, 2016); instead, the prime shift index shows up as significant in our model. We thus reject the claim that the repetition effect is the result of interspeaker variation.[10]

6.2 RQ3: Priming and social factors

Before discussing the effect of gender on priming behavior, it is worth mentioning that women were more shifted (i.e., further along in the NZE SFVS) than men overall (see Figure 6), with the largest difference between men and women in the TRAP vowel. This result is consistent with a long line of findings in variationist sociolinguistics showing women leading in changes in progress.

The model included significant interactions between the prime shift index and gender, indicating that priming behavior did not apply uniformly across gender. In particular, the repetition effect was stronger among men than women, although this pattern is more pronounced in some prime–target combinations than others. For example, as Figure 7 shows, men's repetition effect is stronger
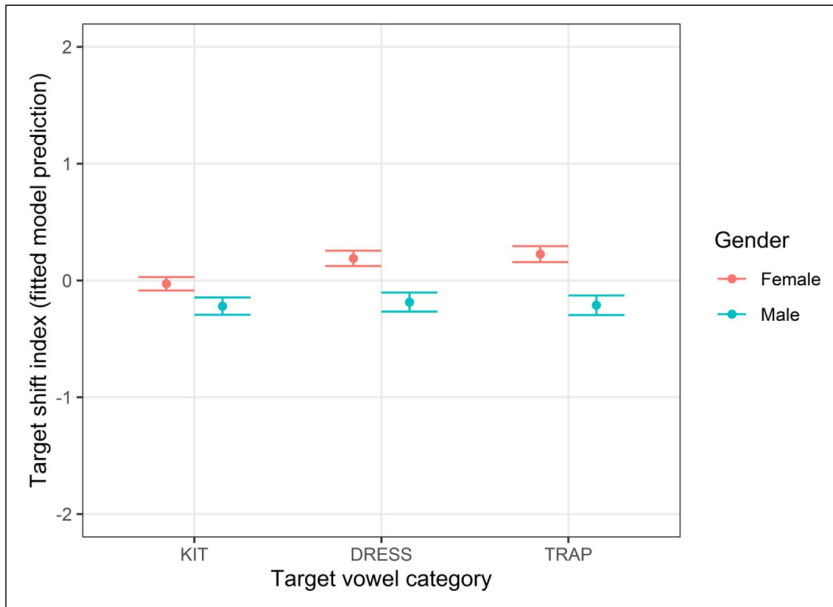
Figure 6. Predicted target shift index by gender, with 95% confidence bars.

than women's with KIT primes and DRESS targets, but men's and women's repetition effects are basically equal with KIT primes and TRAP targets. In Section 7, we entertain competing hypotheses for this surprising gender effect.

We can entertain at least two hypotheses for this surprising gender effect: stylistic cohesion and shift boundedness. The first hypothesis posits that this result is simply caused by speakers' use of similar variants in successive productions reflecting their projection of styles. For this to be the case, we would need to imagine that men in New Zealand project more cohesive styles than women, hence there is less priming among the women (for some vowel combinations) because there is less consistency in shiftedness of prime–target combinations. This account seems unlikely, as the broad claim that men maintain more cohesive styles than women is generally not reflected in the literature on style. Our skepticism aside about this hypothesis aside, it would be supported by a finding that style-shifts are associated with diminished priming behavior.[11]

An alternative hypothesis for this finding that the priming effects are stronger (for some vowel combinations) among men may be related to the fact that women appear to lead men in the SFVS overall (Figure 6; see also Hay et al., 2015). If there are more women than men for whom the shift is nearing completion, then their productions are likely to be less variable than the men's, meaning that they have less room to prime themselves because their primes start out from a more shifted position than men's primes do. This hypothesis would be supported by a finding that younger speakers, who are further advanced in the shift in apparent time (Hay et al., 2015), have smaller repetition effects. Unfortunately, we were unable to directly test this interpretation in this data set given the due-diligence approach to collinearity adopted by our modeling procedure (see Section 4.2), since models that attempted to add age group as a predictor suffered excessive collinearity. Further investigation is needed to assess these hypotheses and shed greater light on the effects driving the interaction of priming and social factors in the present data.
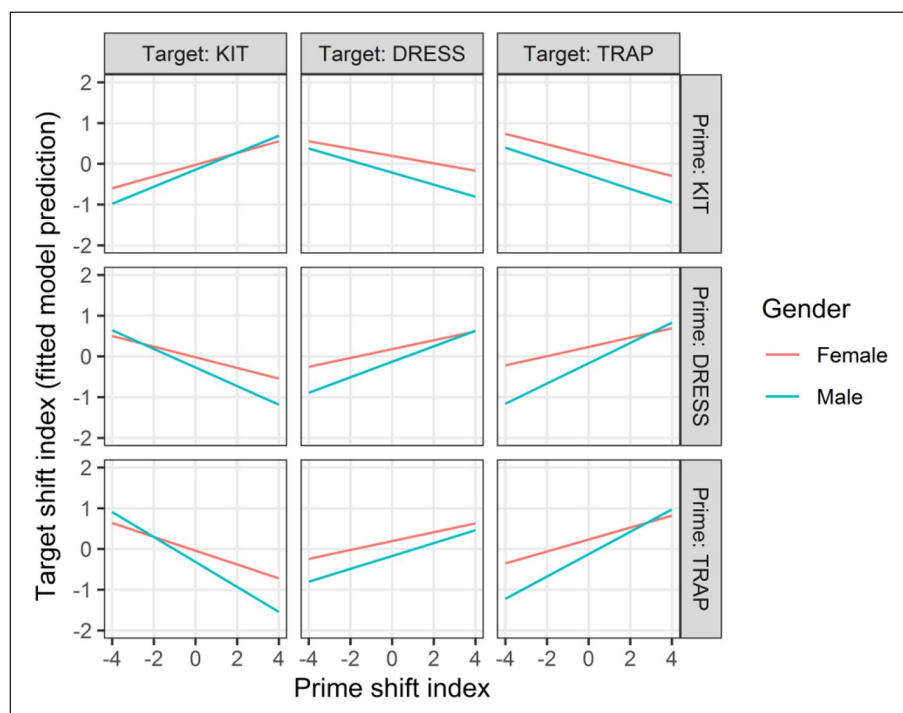
Figure 7. Predicted target shift index by the prime shift index, prime vowel category, target vowel category, and gender.

In summary, these results indicate support for hypothesis 3 (as far as is possible with only one social variable), as we do observe differences in priming behavior for women and men. However, it is not immediately clear why these differences occur.

6.3 Control predictors

The model of the target shift index also included two significant control predictors that were not directly related to any of our hypotheses: speech rate[12] and prime morphemic status. As mentioned above, these predictors were included to control for additional linguistic factors influencing variation in the SFVS. With respect to speech rate, in most prime–target combinations, the repetition effect is diminished with a faster speech rate; this pattern is reversed for two prime–target combinations, as for KIT/DRESS primes and DRESS targets, the repetition effect is enhanced with a faster speech rate. Interestingly, this effect acts as a counterweight to the decay effect described above; when speakers are speaking faster, they are also more likely to have less time between the prime and target (recall that we only added predictors whose inclusion did not cause excessive collinearity).

With respect to the morphemic status of the prime word, in most prime–target combinations, the prime exerts a stronger effect on the target when the prime is a lexical morpheme than when the prime is a grammatical morpheme (Figure 8). This effect is strongest for TRAP–TRAP and TRAP–DRESS prime–target combinations, and potentially absent for KIT–TRAP, DRESS–TRAP, and DRESS–DRESS. This pattern may be related to the different distributions of the target morpheme and prime morpheme by target vowel category in the final model (Table 4); DRESS targets were far more likely than KIT/
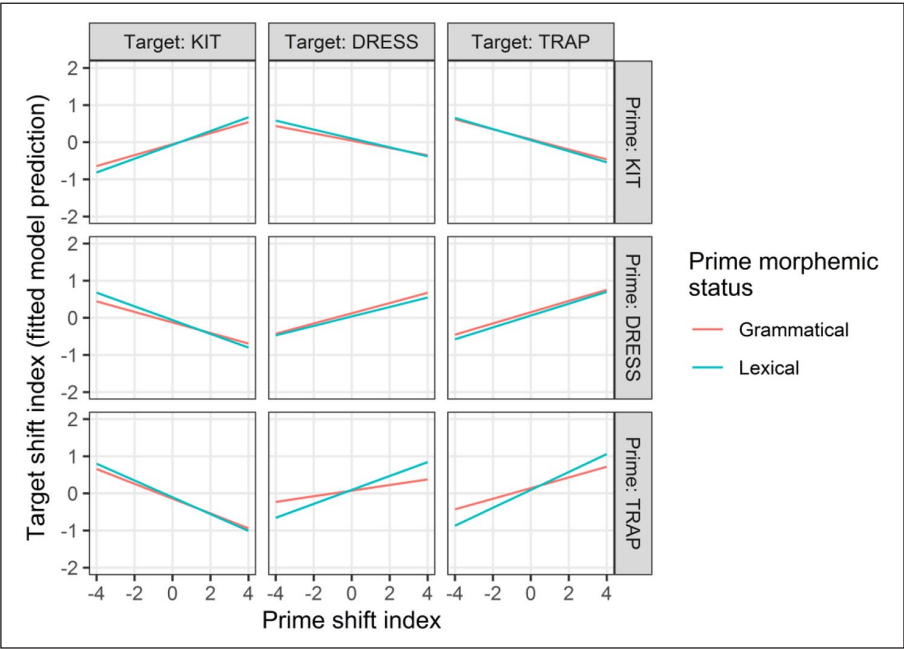
Figure 8. Predicted target shift index by the prime shift index, target vowel category, prime vowel category, and prime morphemic status.

Table 4. Distribution of the target morpheme and prime morpheme by target vowel category among final model data. Since the prime morpheme was included as a predictor and morphemes of ambiguous type do not form a unified class, the final model excluded tokens whose prime morpheme was ambiguous.

Target: KIT		Target morpheme		
		Grammatical	Lexical	Ambiguous
Prime morpheme	Grammatical	9888	5704	0
	Lexical	7250	4342	0

Target: DRESS		Target morpheme		
		Grammatical	Lexical	Ambiguous
Prime morpheme	Grammatical	1835	6148	0
	Lexical	1275	4463	0

Target: TRAP		Target morpheme		
		Grammatical	Lexical	Ambiguous
Prime morpheme	Grammatical	5411	3124	1956
	Lexical	4827	2237	1364

TRAP targets to be lexical morphemes. A full listing of grammatical morphemes by target vowel category can be found in Appendix B.[13]

7 Discussion

Previous work exploring repetition or recency effects in corpora of natural talk have taken a variable-centric approach and asked: does a speaker's previous realization influence their following

realization for a given variable feature in language? This previous work has largely shown that repetition effects that occur in natural speech in real time are strikingly similar to those found in experimental studies of priming. However, while these findings represent a useful first step in broadening our understanding of whether and how priming might operate to constrain variation in natural speech, they are limited in their focus on single variables in isolation; we argue that a single-variable approach to priming misses an important part of the picture given strong empirical evidence that variables do not behave independently of each other, either because they form part of a speech style or because they are structurally related, as in the case of vowel shifts.

While holding constant a number of known social and linguistic predictors of variation on the NZE SFVS, and including random effects of word and speaker, our model revealed a repetition effect such that the phonetic shape of each instance of the NZE SFVS could be predicted in part by the phonetic shape of the previous instance. Importantly, this repetition effect also interacted with both the time difference between the prime and the target (i.e., decay) and the repetition of the same lexical item in the prime and target (i.e., lexical boost) in ways that are to be expected if this repetition effect is at least partially motived by priming.

This repetition effect occurred both within vowel categories and across vowel categories in the SFVS, such that the phonetic shape of (for example) an instance of TRAP could be predicted in part by the phonetic shape of a preceding TRAP, DRESS, or KIT token (whichever was most recent). However, the *direction* of the relationship ran counter to our expectations. We hypothesized that a prime that was more shifted with respect to the SFVS would encourage its target to also be more shifted—that is, that the priming effect would follow the same direction as the change in progress. In reality, the repetition effect we see here seems not to operate in this way; rather, a peripheral prime encourages a peripheral target, including for prime–target combinations for which the prime and target participate in opposing changes in progress (TRAP/DRESS raising/fronting versus KIT centralization). At face value, this priming effect seems to contradict the community-level change over time that the SFVS represents; how can a speaker's previous high front TRAP prime the next instance of KIT to be high and front, while at the community level, TRAP has been raising and fronting at the same time that KIT has been centralizing? One thing to consider is that the time-frames under investigation in this paper are at the micro level—we are talking about momentary, fleeting, utterance-by-utterance influences. It is not immediately clear yet how these micro-level clusters of phonetic variation feed into and ultimately become macro clusters of phonetic variation and change that we see as whole systems of language change (such as vowel shifts) operating over tens if not hundreds of years. It is also worth highlighting that we have shown that the phonetic shape of the prime merely *influences*—not *determines*—that of the target. As shown by our complex multivariate statistical model, a great many different constraints compete to influence the realization of these vowels. Indeed, priming cannot be the most important predictor of variation because, if it were, we would see much less variation than we actually do in natural language. Most models of language variation and change assume no relationship between instances of variation in speech; the present research adds to a body of previous findings (e.g., Clark, 2018; Tamminga, 2016) suggesting that in making this assumption, these models are, at best, impoverished.

A social factor, gender, was found to influence the overall production of the NZE SFVS in expected ways, with women leading the shift. More importantly, gender also interacts with priming effects, suggesting that priming does not necessarily apply uniformly across all social groups. There is, of course, no reason to expect that priming should be different across different social groups and so the existing psycholinguistic literature on individual speaker differences in priming does not consider this (as discussed above, the argument is often that differences in working memory or, depending on the priming task, reading strategy, account for these differences). However, in the case of a sound change in progress that is socially conditioned in a community, it may well

be the case that the social and the cognitive constraints operating on that sound change interact with one another. Again, in line with Clark (2018), this paper suggests that there may be good reason to suspect that social factors can and do constrain priming behavior.

This study also has methodological implications for research in language variation and change, arguing in favor of often-overlooked phenomena and data. Ours is hardly the first variationist study finding a priming effect (Clark, 2014, 2018; Poplack, 1980; Scherre & Naro, 1991, 1992; Tamminga, 2016, 2019; Travis, 2007; Weiner & Labov, 1983), although it is unique in examining priming in continuous phonetic variation. Given that accounting for the factors influencing language variation and change is a central pursuit of variationist theory (i.e., the *constraints* question of Weinreich et al., 1968), this body of work strongly suggests that variationist researchers should at least attempt to include priming predictors as control variables in their statistical models (we say "attempt" because doing so may not always be statistically feasible). In addition, this study's approach to vowel data diverged from typical sociophonetic practice in two ways (see Section 4.1). We declined to exclude tokens that are usually excluded on suspicion of being "ill-behaving" (unstressed vowels, vowels in grammatical morphemes), and we measured vowel shifts along tilted axes using vowel-specific combinations of F1 and F2, in contrast to the usual approach of modeling F1 and F2 separately (e.g., Hay et al., 2015) or Labov et al.'s (2013) approach of using the same tilted axes for multiple variables. In particular, the inclusion of as many tokens as possible allowed us to detect a difference in the strength of the repetition effect according to prime morphemic status. We suggest that future sociophonetic research should be more open to including tokens in grammatical morphemes so future research can better illuminate the role that grammatical morphemes play in vowel change.[14]

7.1 What underlies these phenomena?

It remains an open question which mechanism(s) underlie the findings we have presented here. We can think of at least three possible accounts, which we label "articulatory," "psycholinguistic," and "socio-stylistic." None are sufficient on their own to fully explain the phenomena we have presented here, and, following Tamminga (2019), there is reason to believe these mechanisms are not mutually exclusive. We have already discussed and rejected a fourth possible explanation—that the repetition effect is an artefact of interspeaker variation—on the grounds that any baseline interspeaker variation is accounted for by the random intercept of speaker in our model.

The "articulatory" account explains these findings as a function of articulatory setting (e.g., jaw height): within a given stretch of discourse, speakers adopt a baseline jaw setting that is relatively higher or lower than their overall jaw setting, meaning two consecutive SFVS tokens are likely to be produced with the same jaw setting and therefore a similar height. The finding that most supports this account is the particular patterns of priming across vowel categories, with high and front TRAP/DRESS priming high and front KIT (and vice versa), suggesting that speakers move between periods of higher jaw settings and lower jaw settings regardless of vowel phonemes. Undermining the articulatory account, however, is the lexical boost effect, which shows that word identity enhances the repetition effect; there is no reason why this should be the case if the repetition effect that we see here is purely articulatory. In addition, articulatory setting is generally considered a stable property of languages and dialects, not subject to variation in a moment-by-moment way over the span of a discourse (e.g., Esling & Wong, 1983; Honikman, 1964; Laver, 1978; Mennen et al., 2010). It has been suggested that speakers may adopt different articulatory settings in different speech styles (Pratt & D'Onofrio, 2017; Ramanarayanan et al., 2013), but an account of the current data that combines the articulatory and socio-stylistic accounts still suffers from the shortcomings of the latter (outlined above).[15]

The "psycholinguistic" account explains the findings in terms of mechanisms that are usually labeled "priming" (in the specific psycholinguistic sense of the term "priming") as opposed to "recency" or "persistence." One such mechanism relies on spreading activation. When speakers use a high and front SFVS prime, it increases the resting activation level of exemplars with a matching acoustic representation, within the exemplar distribution for that word (with this additional activation for the prime decaying over time); as a result, when the speaker retrieves a target, they are more likely to retrieve an exemplar similar to that of the prime. In this account, a word match between the prime and target also increases the likelihood of retrieving a target similar to the prime, because the target is being retrieved from the exemplar distribution corresponding to the most recently activated lexeme. Most crucially, the decay that we see is highly indicative of spreading activation, so it is really the decay effect that most clearly adds weight to our interpretation of these results as priming.[16] The "psycholinguistic" account thus provides a natural explanation for the effects found in our data: repetition, lexical boost, and—perhaps the effect that most closely resembles those found in the priming literature more generally—decay. The cross-category priming effects in our data are also reminiscent of the phenomenon reported by Travis et al. (2017), wherein within-language priming of variable subject expression was stronger than cross-language priming. A similar mechanism could perhaps apply to vowel variables coexisting in a monolingual speech community, for example if the speech community considers variation in TRAP and DRESS to be related to one another but considers KIT variation to be distinct (although this is an empirical question not addressable within our data).[17] However, this account alone fails to explain the finding that the repetition effect was stronger for men than women, as there is no conceivable reason that men should be more prone to spreading activation or to the retrieval of recently activated exemplars. It is for this reason that theories of priming generally do not posit that social factors should affect priming behavior, instead explaining variation in priming behavior in terms of individual cognitive differences, such as in working memory capacity, which do not translate to macro-social groups (e.g., Andrews & Lo, 2013; Medeiros & Duñabeitia, 2016; Ortells et al., 2016).

The "socio-stylistic" account explains these findings in terms of the construction and maintenance of sociolinguistic styles and registers (e.g., Eckert, 2000). Similar phonetic variants of socially meaningful variables exhibit temporal clustering as speakers project cohesive styles, with covariation among variants of different variables that are bound together through bricolage (e.g., Eckert, 2008), enregisterment (e.g., Agha, 2004), or some other means of conveying social meaning through covariation. As speakers move through the discourse, they agentively alter their variation to reflect and constitute different topics, stances, styles, and levels of formality (Eckert, 2018). While the socio-stylistic account straightforwardly explains the repetition effect, it fails to adequately explain the other core priming effects: decay and lexical boost. While a longer prime target time difference means a greater likelihood of a switch in styles, registers, etc., between the prime and target, the plausible time-scale of such an effect conflicts with our findings; it seems implausible that there should be much difference between a 20 millisecond delay and a 670 millisecond delay in terms of switching styles. Likewise, the socio-stylistic account cannot explain the lexical boost effect. Although individual lexical items may be suffused with particularly strong indexical meanings, such as the intensifier/determiner *hella* triggering an association with Northern California (Bucholtz et al., 2007), the lexemes in our data for which the target word matched the prime word were overwhelmingly function words (e.g., *that*, *it*, *is*, *and*) that are unlikely to be the site of lexically specified social meaning-making. Finally, while the socio-stylistic account focuses on covariation of different variants, it likely would not predict the specific patterns of priming across vowel categories found here. Given the historical association of TRAP raising/fronting, DRESS raising/fronting, and KIT centralization that make up the SFVS characteristic of NZE, priming effects driven by stylistic cohesion would likely predict a relationship between high front TRAP/

DRESS and central KIT (and vice versa); instead, we find a relationship between high front TRAP/DRESS and high front KIT. In short, while there is no reason to doubt the importance of style in driving patterns of variation in unfolding discourse—as has amply been demonstrated by a large body of work in third-wave variationist sociolinguistics (Eckert, 2018)—it is clear to us that socio-stylistic factors are inadequate to account for "sequential dependence" (Tamminga, 2019) specifically or priming phenomena generally as they are defined here.

8 Conclusion

Although the driving mechanism for these findings remains an open question, we have demonstrated that continuous phonetic variation is indeed subject to intraspeaker priming, exhibiting repetition, decay (albeit on a shorter time-scale), and lexical boost effects that are characteristic signatures of priming. We also find that priming takes place across vowel categories that are interrelated by their membership in the NZE SFVS, although in a direction that potentially suggests articulatory motivations (peripherality) rather than social motivations (chain shifting and indexical linkages). Contrary to typical theories of priming that are predicated on spreading activation and assume only individual differences in priming behavior, we find evidence for differential priming behavior between women and men. In summary, the present research adds to a growing body of literature highlighting the important role that intraspeaker priming plays in shaping sociolinguistic variation.

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Notes

1. Medial /t/ is currently undergoing change in NZE with a shift away from canonical [t]. There are several phonetic realizations of this variable but most studies (see, e.g., Hay & Foulkes, 2016) collapse the voiceless variants together and the voiced variants together. Clark (2018) adopted the same methodology.
2. An anonymous reviewer questions whether this is the same as the lexical boost phenomenon found in experimental studies of syntactic priming (e.g., Hartsuiker et al., 2008; Jaeger & Snider, 2013). Because our study uses a corpus rather than experimental data, we are not in a position to answer this question definitively, and we use the term *lexical boost* as a description of the pattern in the data rather than an explanation implying an underlying mechanism.

3. In the notation for English lexical sets introduced by Wells (1982), TRAP, DRESS, and KIT refer to the vowel categories canonically represented by the IPA symbols /æ, ε, ɪ/—but without making claims about the phonological state of these classes' "underlying representations."

4. In fact, this considerable intraspeaker variation is not only limited to the SFVS—in recent work by the second author, we find that other vowel changes co-vary with this chain shift (Brand et al., 2019; see also the longer under-review manuscript at https://osf.io/q4j29/).

5. These were the same narratives as in Clark's (2018) medial (t) study, save for one speaker whose data was removed because his vowel measurements were consistently inaccurate.

6. Out of 91,146 short front vowel tokens in the original data set, 7945 were removed as outliers (7120 detected automatically, 825 via hand-checking). This subset of 83,201 non-outlier tokens was the basis for calculating the shift index. Of the remaining tokens, another 155 were removed for being the speaker's first token; another 7129 were removed because they followed outliers; another 4920 were removed for being in the same word as their prime; another 7566 were removed because of prime target time difference measurements; and another 3607 were removed for having ambiguous prime morphemic status. This yielded the 59,824 tokens used for analysis.

7. Collinearity was assessed on the basis of the generalized variance inflation factor (GVIF) via the `vif()` function in the `car` R package (Fox & Monette, 1992; Fox & Weisberg, 2011), using the usual "less than 10" rule of thumb for variance inflation factors (O'Brien, 2007).Readers should note that the GVIF returned by `vif()` needs to be squared before comparing to 10 or any other rule of thumb. Fox and Weisberg (2011, p. 325) note, "if there are p regressors in a term, then $GVIF^{1/2p}$ is a one-dimensional expression of the decrease in the precision of estimation due to collinearity—analogous to taking the square root of the usual VIF. When $p = 1$, the GVIF reduces to the usual VIF."

8. The remaining predictors were all excluded from the model because adding them would have yielded excessive collinearity in the model. In other words, it is possible that some effects that actually influence these vowels are nevertheless excluded from a model specifically constructed to test hypotheses about priming because their inclusion causes the model to be untenable. Again, we have made the data for this analysis available at https://github.com/nzilbb/Priming-SFVS-Data so readers can test how the modeling pans out differently under assumptions other than those that guided our modeling procedure.

9. Among the 2328 tokens for which the target word was a repetition of the prime word, all but 1.5% had the same vowel category for the prime and target; as a result, Figure 5 displays effects only for within-vowel priming. The few exceptions were words with multiple SFVS vowels, such as *everything* or *liquefaction*.

10. We should note that this is the case in the variables we studied, which are subject to well-documented fine-grained phonetic variation, but would not necessarily hold across all variables in the spontaneous speech/personal narrative context from which we drew our data. Grammatical/discourse variables in particular are subject to being used in the construction of idiosyncratic speech styles in spontaneous speech (e.g., Johnstone, 1996; van Compernolle & Williams, 2012; Villarreal, 2014)—while also being subject to priming effects in experimental settings, as discussed in Section 1—and so could be candidates for studying where automaticity ends and style begins.

11. The data in the present paper are not well suited to testing this hypothesis because the similarity in discourse situations across speakers limits the stylistic variation in the data.

12. The local speech rate was calculated as syllables per second within a breath group, excluding all inter-word pauses longer than 50 ms long. The speech rate predictor in our model controlled for interspeaker speech rate differences by dividing the local speech rate by the speaker's mean local speech rate.

13. An anonymous reviewer suggests that the weaker effects of grammatical prime TRAP are related to the prevalence of *that* and *and* among the TRAP tokens. This may be the case, although it is unclear why this effect would not be replicated for prime TRAP and target KIT (Figure 8), nor why the TRAP grammatical morphemes would be substantially weaker than high-frequency KIT grammatical morphemes that include *-ing*, *it*, and *in*.

14. Thanks to an anonymous reviewer for making this point.

15. Two anonymous reviewers suggest that the articulatory account would be strengthened by a finding that SFVS vowels show cross-vowel priming with "a vowel that is diachronically stable and not (as far as

we know) socio-stylistically linked to the chain shift"—in other words a control vowel. We wanted to perform this analysis, but it turns out that there are basically no suitable control vowels in this speech community; Brand et al. (2019) show that in NZE, other vowels co-vary with the SFVS and there are basically no diachronically stable vowels. This question thus remains open for future work in speech communities in which the diachronic vowel variation is better suited to such an analysis.

16. We thank the Associate Editor for highlighting this point to us in the review.
17. We thank an anonymous reviewer for making this point.

References

Agha, A. (2004). Registers of language. In A. Duranti (Ed.), *A companion to linguistic anthropology* (pp. 23-45). Blackwell.

Akaike, H. (1974). A new look at the statistical model identification. *IEEE Transactions on Automatic Control*, *19*(6), 716-723. https://doi.org/10.1109/TAC.1974.1100705

Andrews, S., & Lo, S. (2013). Is morphological priming stronger for transparent than opaque words? It depends on individual differences in spelling and vocabulary. *Journal of Memory and Language*, *68*(3), 279-296. https://doi.org/10.1016/j.jml.2012.12.001

Babel, M. (2010). Dialect divergence and convergence in New Zealand English. *Language in Society*, *39*(4), 437-456. https://doi.org/10.1017/S0047404510000400

Babel, M. (2012). Evidence for phonetic and social selectivity in spontaneous phonetic imitation. *Journal of Phonetics*, *40*(1), 177-189. https://doi.org/10.1016/j.wocn.2011.09.001

Bell, A. (1984). Language style as audience design. *Language in Society*, *13*(2), 145-204. Available at http://www.jstor.org/stable/4167516

Bell, A. (2001). Back in style: Reworking audience design. In P. Eckert & J. R. Rickford (Eds.), *Style and sociolinguistic variation* (pp. 139-169). Cambridge University Press.

Bock, K. (1986). Syntactic persistence in language production. *Cognitive Psychology*, *18*(3), 355-387. https://doi.org/10.1016/0010-0285(86)90004-6

Bock, K., & Loebell, H. (1990). Framing sentences. *Cognition*, *35*(1), 1-39. https://doi.org/10.1016/0010-0277(90)90035-I

Brand, J., Hay, J., Clark, L., Watson, K., & Sóskuthy, M. (2019). *Systematic co-variation of monophthongs across speakers of New Zealand English*. Paper presented at 19th ICPhS, Melbourne.

Branigan, H. P., Pickering, M. J., & Cleland, A. A. (2000). Syntactic co-ordination in dialogue. *Cognition*, *75*(2), B13-B25. https://doi.org/10.1016/S0010-0277(99)00081-5

Bucholtz, M., Bermudez, N., Fung, V., Edwards, L., & Vargas, R. (2007). Hella Nor Cal or totally So Cal? The perceptual dialectology of California. *Journal of English Linguistics*, *35*(4), 325-352. https://doi.org/10.1177/0075424207307780

Clark, L. (2014). Phonological repetition effects in natural conversation: evidence from TH-fronting in Fife. In R. Lawson (Ed.), *Sociolinguistics in Scotland* (pp. 153-176). Palgrave Macmillan.

Clark, L. (2018). Priming as a motivating factor in sociophonetic variation and change. *Topics in Cognitive Science*, *10*(4), 1-16. https://doi.org/10.1111/tops.12338

Clark, L., MacGougan, H., Hay, J., & Walsh, L. (2016). "Kia ora. This is my earthquake story". Multiple applications of a sociolinguistic corpus. *Ampersand*, *3*, 13-20. https://doi.org/10.1016/j.amper.2016.01.001

Delvaux, V., & Soquet, A. (2007). The influence of ambient speech on adult speech productions through unintentional imitation. *Phonetica*, *64*(2-3), 145-173. https://doi.org/10.1159/000107914

Docherty, G. J., & Foulkes, P. (2014). An evaluation of usage-based approaches to the modelling of sociophonetic variability. *Lingua*, *142*, 42-56. https://doi.org/10.1016/j.lingua.2013.01.011

Eckert, P. (2000). *Linguistic variation as social practice*. Blackwell.

Eckert, P. (2008). Variation and the indexical field. *Journal of Sociolinguistics*, *12*(4), 453-476. https://doi.org/10.1111/j.1467-9841.2008.00374.x

Eckert, P. (2012). Three waves of variation study: The emergence of meaning in the study of sociolinguistic variation. *Annual Review of Anthropology*, *41*(1), 87-100. https://doi.org/10.1146/annurev-anthro-092611-145828

Eckert, P. (2018). *Meaning and linguistic variation: The third wave in sociolinguistics*. Cambridge University Press.

Esling, J. H., & Wong, R. F. (1983). Voice quality settings and the teaching of pronunciation. *TESOL Quarterly*, *17*(1), 89-95. https://doi.org/10.2307/3586426

Fox, J., & Monette, G. (1992). Generalized collinearity diagnostics. *Journal of the American Statistical Association*, *87*(417), 178-183. https://doi.org/10.2307/2290467

Fox, J., & Weisberg, S. (2011). *An R companion to applied regression* (2nd ed.). SAGE.

Giles, H., & Coupland, N. (1991). *Language: Contexts and consequences*. Open University Press.

Giles, H., Coupland, N., & Coupland, J. (1991). Accommodation theory: Communication, context, and consequence. In H. Giles, J. Coupland, & N. Coupland (Eds.), *Contexts of accommodation: Developments in applied sociolinguistics* (pp. 1-68). Cambridge University Press.

Goldinger, S. D., Luce, P. A., Pisoni, D. B., & Marcario, J. K. (1992). Form-based priming in spoken word recognition: The roles of competition and bias. *Journal of Experimental Psychology: Learning, Memory, and Cognition*, *18*(6), 1211-1238. https://www.ncbi.nlm.nih.gov/pubmed/1447548 https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3514873/

Gordon, E., Campbell, L., Hay, J., Maclagan, M., Sudbury, A., & Trudgill, P. (2004). *New Zealand English: Its origins and evolution*. Cambridge University Press.

Gries, S. T. (2005). Syntactic priming: A corpus-based approach. *Journal of Psycholinguistic Research*, *34*(4), 365-399. https://doi.org/10.1007/s10936-005-6139-3

Guy, G. R. (2013). The cognitive coherence of sociolects: How do speakers handle multiple sociolinguistic variables? *Journal of Pragmatics*, *52*, 63-71. https://doi.org/10.1016/j.pragma.2012.12.019

Hartsuiker, R. J., Bernolet, S., Schoonbaert, S., Speybroeck, S., & Vanderelst, D. (2008). Syntactic priming persists while the lexical boost decays: Evidence from written and spoken dialogue. *Journal of Memory and Language*, *58*(2), 214-238. https://doi.org/10.1016/j.jml.2007.07.003

Hay, J., & Foulkes, P. (2016). The evolution of medial /t/ over real and remembered time. *Language*, *92*(2), 298-330. https://doi.org/10.1353/lan.2016.0036

Hay, J., Pierrehumbert, J. B., Walker, A. J., & LaShell, P. (2015). Tracking word frequency effects through 130 years of sound change. *Cognition*, *139*, 83-91. https://doi.org/10.1016/j.cognition.2015.02.012

Honikman, B. (1964). Articulatory settings. In D. Abercrombie, D. B. Fry, P. A. D. MacCarthy, N. C. Scott, & J. L. M. Trim (Eds.), *In honour of Daniel Jones* (pp. 73-84). Longman.

Jaeger, T. F., & Snider, N. E. (2013). Alignment as a consequence of expectation adaptation: Syntactic priming is affected by the prime's prediction error given both prior and recent experience. *Cognition*, *127*(1), 57-83. https://doi.org/10.1016/j.cognition.2012.10.013

Johnstone, B. (1996). *The linguistic individual: Self-expression in language and linguistics*. Oxford University Press.

Kuznetsova, A., Brockhoff, B., & Christensen, H. B. (2016). LmerTest (Version 2.*0-33) [R package]*. https://CRAN.R-project.org/package=lmerTest

Labov, W. (1994). *Principles of linguistic change, vol 1: Internal factors*. Blackwell.

Labov, W. (2001). *Principles of linguistic change, vol. 2: Social factors*. Blackwell.

Labov, W. (2010). *Principles of linguistic change Vol. III: Cognitive and cultural factors*. Wiley-Blackwell.

Labov, W., Ash, S., & Boberg, C. (2006). *The atlas of North American English: Phonetics, phonology and sound change*. Mouton de Gruyter.

Labov, W., Rosenfelder, I., & Freuhwald, J. (2013). One hundred years of sound change in Philadelphia: Linear incrementation, reversal, and reanalysis. *Language*, *89*(1), 30-65. https://doi.org/10.1353/lan.2013.0015

Labov, W., Yaeger, M., & Steiner, R. (1972). *A quantitative study of sound change in progress. Report on National Science Foundation contract NSF-GS-3287*. University of Pennsylvania.

Langstrof, C. (2006). *Vowel change in New Zealand English: Patterns and implications*. (PhD thesis), University of Canterbury.

Laver, J. (1978). The concept of articulatory settings: An historical survey. *Historiographia Linguistica*, *5*(1-2), 1-14. https://doi.org/https://doi.org/10.1075/hl.5.1-2.02lav

Medeiros, J., & Duñabeitia, J. A. (2016). Not everybody sees the ness in the darkness: Individual differences in masked suffix priming. *Frontiers in Psychology*, *7*(1585), 1-10. https://doi.org/10.3389/fpsyg.2016.01585

Mennen, I., Scobbie, J. M., de Leeuw, E., Schaeffler, S., & Schaeffler, F. (2010). Measuring language-specific phonetic settings. *Second Language Research*, *26*(1), 13-41. https://doi.org/10.1177/0267658309337617

O'Brien, R. M. (2007). A caution regarding rules of thumb for variance inflation factors. *Quality & Quantity*, *41*(5), 673-690. https://doi.org/10.1007/s11135-006-9018-6

Ortells, J. J., Noguera, C., Álvarez, D., Carmona, E., & Houghton, G. (2016). Individual differences in working memory capacity modulates semantic negative priming from single prime words. *Frontiers in Psychology*, *7*(1286), 1-14. https://doi.org/10.3389/fpsyg.2016.01286

Poplack, S. (1980). Sometimes I'll start a sentence in Spanish y termino en Español: Toward a typology of code-switching. *Linguistics*, *18*(7/8), 581-618. https://doi.org/10.1515/ling.1980.18.7-8.581

Pratt, T., & D'Onofrio, A. (2017). Jaw setting and the California Vowel Shift in parodic performance. *Language in Society*, *46*(3), 283-312. https://doi.org/10.1017/S0047404517000227

R Core Team. (2018). R: A language and environment for statistical computing (Version 3.5.2). *Vienna*. https://www.R-project.org/

Ramanarayanan, V., Goldstein, L., Byrd, D., & Narayanan, S. S. (2013). An investigation of articulatory setting using real-time magnetic resonance imaging. *The Journal of the Acoustical Society of America*, *134*(1), 510-519. https://doi.org/10.1121/1.4807639

Satterthwaite, F. E. (1946). An approximate distribution of estimates of variance components. *Biometrics Bulletin*, *2*(6), 110-114. https://doi.org/10.2307/3002019

Scherre, M. M. P., & Naro, A. J. (1991). Marking in discourse: "Birds of a feather". *Language Variation and Change*, *3*(1), 23-32. https://doi.org/10.1017/S0954394500000430

Scherre, M. M. P., & Naro, A. J. (1992). The serial effect on internal and external variables. *Language Variation and Change*, *4*(1), 1-13. https://doi.org/10.1017/S0954394500000636

Sumner, M., Kim, S. K., King, E., & McGowan, K. (2014). The socially weighted encoding of spoken words: A dual-route approach to speech perception. *Frontiers in Psychology*, *4*(1015). https://doi.org/10.3389/fpsyg.2013.01015

Szmrecsanyi, B. (2005). Language users as creatures of habit: A corpus-linguistic analysis of persistence in spoken English. *Corpus Linguistics and Linguistic Theory*, *1*(1), 113-150. https://doi.org/10.1515/cllt.2005.1.1.113

Szmrecsanyi, B. (2006). *Morphosyntactic persistence in spoken English: A corpus study at the intersection of variationist sociolinguistics, psycholinguistics, and discourse analysis*. Mouton de Gruyter.

Tamminga, M. (2014). *Persistence in the production of linguistic variation*. (Ph.D. dissertation). University of Pennsylvania. https://repository.upenn.edu/edissertations/1466/

Tamminga, M. (2016). Persistence in phonological and morphological variation. *Language Variation and Change*, *28*(3), 335-356. https://doi.org/10.1017/S0954394516000119

Tamminga, M. (2019). Sources of microtemporal clustering in sociolinguistic sequences. *Frontiers in Artificial Intelligence*, *2*(10), 1-17. https://doi.org/10.3389/frai.2019.00010

Tamminga, M., Ahern, C., & Ecay, A. (2016). Generalized Additive Mixed Models for intraspeaker variation. *Linguistics Vanguard*, *2*(s1), 33-41. https://doi.org/10.1515/lingvan-2016-0030

Travis, C. E. (2007). Genre effects on subject expression in Spanish: Priming in narrative and conversation. *Language Variation and Change*, *19*(2), 101-135. https://doi.org/10.1017/s0954394507070081

Travis, C. E., Torres Cacoullos, R., & Kidd, E. (2017). Cross-language priming: A view from bilingual speech. *Bilingualism: Language and Cognition*, *20*(2), 283-298. https://doi.org/10.1017/S1366728915000127

van Compernolle, R. A., & Williams, L. (2012). Reconceptualizing sociolinguistic competence as mediated action: Identity, meaning-making, agency. *Modern Language Journal*, *96*(2), 234-250. https://doi.org/10.1111/j.1540-4781.2012.01334.x

Villarreal, D. (2014). Connecting production to judgments: T/V address forms and the L2 identities of intermediate Spanish learners. *Journal of Pragmatics*, *66*, 1-14. https://doi.org/10.1016/j.pragma.2014.02.005

Weiner, E. J., & Labov, W. (1983). Constraints on the agentless passive. *Journal of Linguistics*, *19*(1), 29-58. https://doi.org/10.1017/S0022226700007441

Weinreich, U., Labov, W., & Herzog, M. I. (1968). Empirical foundations for a theory of language change. In W. P. Lehmann & Y. Malkiel (Eds.), *Directions for historical linguistics* (pp. 97-195). University of Texas Press.

Wells, J. C. (1982). *Accents of English*. Cambridge University Press.

Appendices

Appendix A. Summary of model fixed effects

In the following table of fixed effects, stars refer to significance levels: $. p < .1$, $*p < .05$, $**p < .01$, $***p < .001$

	β	SE	df	t	p
(Intercept)	−0.01384	0.03009	599	−0.460	0.645671
PrimeShiftIndex	0.12473	0.01064	1548	11.728	< 2e-16***
Target DRESS	0.18078	0.03483	4678	5.190	2.19e-07***
Target TRAP	0.24466	0.03600	4498	6.797	1.21e-11***
Prime DRESS	−0.03286	0.02711	57725	−1.212	0.225500
Prime TRAP	−0.04102	0.01672	52876	−2.453	0.014154*
scale(PrimeTargetTimeDiff)	−0.00978	0.00688	58578	−1.420	0.155530
GenderM	−0.12096	0.04044	206	−2.991	0.003118**
scale(SpeechRateDev)	0.02744	0.00674	58652	4.073	4.64e-05***
PrimeMorphemeLexical	−0.01937	0.01445	58599	−1.341	0.180067
PrimeTargetSameWordTRUE	−0.04822	0.02443	57062	−1.974	0.048437*
PrimeShiftIndex:Target DRESS	−0.21219	0.01755	4185	−12.090	< 2e-16***
PrimeShiftIndex:Target TRAP	−0.25056	0.01668	1397	−15.023	< 2e-16***
Prime DRESS:scale(PrimeTargetTimeDiff)	0.00249	0.01238	58244	0.201	0.840581
Prime TRAP:scale(PrimeTargetTimeDiff)	0.02136	0.01183	58383	1.806	0.070846.
Prime DRESS:GenderM	−0.12935	0.02758	32065	−4.690	2.74e-06***
Prime TRAP:GenderM	−0.15734	0.02652	32189	−5.933	3.01e-09***
Prime DRESS:scale(SpeechRateDev)	−0.02635	0.01228	58252	−2.147	0.031831*
Prime TRAP:scale(SpeechRateDev)	−0.01161	0.01182	58516	−0.982	0.326033
Prime DRESS:PrimeMorphemeLexical	0.08216	0.03090	58361	2.659	0.007836**
Prime TRAP:PrimeMorphemeLexical	0.05517	0.02525	58467	2.185	0.028910*
PrimeShiftIndex:Prime DRESS	−0.24046	0.02408	55217	−9.984	< 2e-16***
PrimeShiftIndex:Prime TRAP	−0.28643	0.01661	52877	−17.247	< 2e-16***
PrimeShiftIndex:scale(PrimeTargetTimeDiff)	−0.05308	0.00689	48359	−7.698	1.41e-14***
PrimeShiftIndex:GenderM	0.06448	0.01487	966	4.335	1.61e-05***
PrimeShiftIndex:scale(SpeechRateDev)	−0.01862	0.00667	55359	−2.792	0.005232**
PrimeShiftIndex:PrimeMorphemeLexical	0.03776	0.01540	56689	2.452	0.014213*
PrimeShiftIndex:PrimeTargetSameWord TRUE	0.08667	0.02567	53359	3.377	0.000735***
Target DRESS:Prime DRESS	0.08410	0.04272	58173	1.969	0.049014*
Target TRAP:Prime DRESS	0.07704	0.04169	57908	1.848	0.064657.
Target DRESS:Prime TRAP	0.05900	0.02823	58463	2.090	0.036604*
Target TRAP:Prime TRAP	0.06485	0.02630	58213	2.466	0.013672*

(Continued)

	β	SE	df	t	p
TargetDRESS:scale(PrimeTargetTimeDiff)	−0.00164	0.01233	58925	−0.133	0.894229
TargetTRAP:scale(PrimeTargetTimeDiff)	0.02558	0.01074	58379	2.381	0.017290*
TargetDRESS:GenderM	−0.28657	0.02653	59100	−10.800	< 2e-16***
TargetTRAP:GenderM	−0.37451	0.02302	58496	−16.272	< 2e-16***
TargetDRESS:scale(SpeechRateDev)	−0.08366	0.01209	58850	−6.919	4.58e-12***
TargetTRAP:scale(SpeechRateDev)	−0.05898	0.01069	58433	−5.519	3.43e-08***
TargetDRESS:PrimeMorphemeLexical	0.07825	0.02643	59005	2.960	0.003074**
TargetTRAP:PrimeMorphemeLexical	−0.00071	0.02255	58384	−0.031	0.975048
TargetDRESS:PrimeTargetSameWordTRUE	0.03360	0.04866	57010	0.691	0.489858
TargetTRAP:PrimeTargetSameWordTRUE	−0.02178	0.04137	56980	−0.527	0.598538
PrimeShiftIndex:PrimeDRESS:scale(Prime TargetTimeDiff)	0.06352	0.01204	56327	5.275	1.33e-07***
PrimeShiftIndex:PrimeTRAP:scale (PrimeTargetTimeDiff)	0.08316	0.01164	56846	7.147	8.98e-13***
PrimeShiftIndex:PrimeDRESS:GenderM	−0.16195	0.02729	51120	−5.935	2.95e-09***
PrimeShiftIndex:PrimeTRAP:GenderM	−0.20090	0.02639	49601	−7.612	2.76e-14***
PrimeShiftIndex:PrimeDRESS:scale(Speech RateDev)	0.01986	0.01227	57688	1.618	0.105662
PrimeShiftIndex:PrimeTRAP:scale(Speech RateDev)	0.03469	0.01157	58257	2.998	0.002720**
PrimeShiftIndex:PrimeDRESS:Prime MorphemeLexical	−0.08044	0.02853	56874	−2.820	0.004804**
PrimeShiftIndex:PrimeTRAP:Prime MorphemeLexical	−0.06409	0.02653	57427	−2.416	0.015706*
TargetDRESS:PrimeDRESS:scale(PrimeTarget TimeDiff)	0.01539	0.02108	58455	0.730	0.465290
TargetTRAP:PrimeDRESS:scale(PrimeTarget TimeDiff)	−0.00587	0.01926	58074	−0.305	0.760294
TargetDRESS:PrimeTRAP:scale(PrimeTarget TimeDiff)	−0.03297	0.02033	58565	−1.622	0.104860
TargetTRAP:PrimeTRAP:scale(PrimeTarget TimeDiff)	−0.05594	0.01798	58156	−3.112	0.001859**
TargetDRESS:PrimeDRESS:GenderM	0.22479	0.04646	58585	4.839	1.31e-06***
TargetTRAP:PrimeDRESS:GenderM	0.22416	0.04292	58191	5.223	1.77e-07***
TargetDRESS:PrimeTRAP:GenderM	0.20055	0.04567	58725	4.391	1.13e-05***
TargetTRAP:PrimeTRAP:GenderM	0.29599	0.04044	58274	7.320	2.52e-13***
TargetDRESS:PrimeDRESS:scale(SpeechRate Dev)	0.03329	0.02070	58319	1.608	0.107867
TargetTRAP:PrimeDRESS:scale(SpeechRate Dev)	0.03350	0.01916	57969	1.748	0.080485.
TargetDRESS:PrimeTRAP:scale(SpeechRate Dev)	0.03648	0.02028	58625	1.798	0.072147.
TargetTRAP:PrimeTRAP:scale(SpeechRateDev)	0.03069	0.01820	58146	1.686	0.091785.
TargetDRESS:PrimeDRESS:PrimeMorpheme Lexical	−0.22713	0.04994	58399	−4.548	5.43e-06***
TargetTRAP:PrimeDRESS:PrimeMorpheme Lexical	−0.15123	0.04762	58014	−3.176	0.001497**

(Continued)

	β	SE	df	t	p
TargetDRESS:PrimeTRAP:PrimeMorpheme Lexical	−0.09449	0.04465	58737	−2.116	0.034323*
TargetTRAP:PrimeTRAP:PrimeMorpheme Lexical	−0.08399	0.03885	58319	−2.162	0.030645*
PrimeShiftIndex:TargetDRESS:PrimeDRESS	0.43446	0.03843	53983	11.306	< 2e-16***
PrimeShiftIndex:TargetTRAP:PrimeDRESS	0.47338	0.03728	57213	12.697	< 2e-16***
PrimeShiftIndex:TargetDRESS:PrimeTRAP	0.42834	0.02722	52748	15.738	< 2e-16***
PrimeShiftIndex:TargetTRAP:PrimeTRAP	0.51364	0.02589	54329	19.841	< 2e-16***
PrimeShiftIndex:TargetDRESS:scale(Prime TargetTimeDiff)	0.07540	0.01235	48512	6.107	1.02e-09***
PrimeShiftIndex:TargetTRAP:scale (PrimeTargetTimeDiff)	0.07709	0.01063	52047	7.254	4.10e-13***
PrimeShiftIndex:TargetDRESS:GenderM	−0.12160	0.02527	56749	−4.812	1.49e-06***
PrimeShiftIndex:TargetTRAP:GenderM	−0.10390	0.02226	55908	−4.669	3.04e-06***
PrimeShiftIndex:TargetDRESS:scale(Speech RateDev)	0.01296	0.01178	56224	1.099	0.271612
PrimeShiftIndex:TargetTRAP:scale(Speech RateDev)	0.02662	0.01069	55339	2.491	0.012744*
PrimeShiftIndex:TargetDRESS:Prime MorphemeLexical	−0.05923	0.02813	57175	−2.106	0.035232*
PrimeShiftIndex:TargetTRAP:Prime MorphemeLexical	−0.05151	0.02432	57176	−2.118	0.034177*
PrimeShiftIndex:TargetDRESS:PrimeTarget SameWordTRUE	0.07504	0.04888	53727	1.535	0.124748
PrimeShiftIndex:TargetTRAP:PrimeTarget SameWordTRUE	−0.01418	0.04193	49676	−0.338	0.735284
PrimeShiftIndex:TargetDRESS:PrimeDRESS: scale(PrimeTargetTimeDiff)	−0.10204	0.02079	55316	−4.907	9.26e-07***
PrimeShiftIndex:TargetTRAP:PrimeDRESS: scale(PrimeTargetTimeDiff)	−0.08271	0.01880	57043	−4.398	1.09e-05***
PrimeShiftIndex:TargetDRESS:PrimeTRAP: scale(PrimeTargetTimeDiff)	−0.11659	0.01983	54497	−5.879	4.15e-09***
PrimeShiftIndex:TargetTRAP:PrimeTRAP:scale (PrimeTargetTimeDiff)	−0.11508	0.01775	57638	−6.483	9.09e-11***
PrimeShiftIndex:TargetDRESS:PrimeDRESS: GenderM	0.30077	0.04591	57878	6.552	5.74e-11***
PrimeShiftIndex:TargetTRAP:PrimeDRESS: GenderM	0.33637	0.04220	57768	7.972	1.59e-15***
PrimeShiftIndex:TargetDRESS:PrimeTRAP: GenderM	0.30701	0.04422	57646	6.943	3.89e-12***
PrimeShiftIndex:TargetTRAP:PrimeTRAP: GenderM	0.36837	0.03929	57903	9.377	< 2e-16***
PrimeShiftIndex:TargetDRESS:PrimeDRESS: scale(SpeechRateDev)	−0.00257	0.02066	57677	−0.124	0.901145
PrimeShiftIndex:TargetTRAP:PrimeDRESS:scale (SpeechRateDev)	−0.03222	0.01909	57634	−1.687	0.091529.
PrimeShiftIndex:TargetDRESS:PrimeTRAP: scale(SpeechRateDev)	−0.05162	0.01969	57378	−2.622	0.008756**

(Continued)

	β	SE	df	t	p
PrimeShiftIndex:TargetTRAP:PrimeTRAP:scale (SpeechRateDev)	−0.06832	0.01800	57875	−3.795	0.000148***
PrimeShiftIndex:TargetDRESS:PrimeDRESS: PrimeMorphemeLexical	0.09172	0.04755	54330	1.929	0.053775.
PrimeShiftIndex:TargetTRAP:PrimeDRESS: PrimeMorphemeLexical	0.10295	0.04465	57574	2.306	0.021128*
PrimeShiftIndex:TargetDRESS:PrimeTRAP: PrimeMorphemeLexical	0.19718	0.04646	57506	4.244	2.20e-05***
PrimeShiftIndex:TargetTRAP:PrimeTRAP: PrimeMorphemeLexical	0.17554	0.04082	56720	4.301	1.70e-05***

Appendix B. Grammatical morphemes by vowel category

KIT

-est (56), *-ing* (4012), *did* (871), *except* (19), *him* (246), *his* (251), *if* (349), *in* (3468), *inside* (51), *into* (338), *is* (816), *it* (4043), *its* (34), *itself* (22), *since* (66), *this* (782), *till* (35), *until* (85), *which* (475), *will* (49), *with* (998), *within* (38), *without* (34)

DRESS

any (123), *anybody* (46), *anyone* (31), *anything* (141), *every* (91), *everybody* (120), *everyone* (158), *everything* (225), *herself* (4), *however* (20), *let* (89), *many* (100), *myself* (66), *ourselves* (13), *several* (8), *them* (357), *themself* (2), *themselves* (16), *then* (892), *unless* (10), *whatever* (49), *when* (463), *whenever* (3), *wherever* (3), *whether* (61), *whoever* (7), *yourself* (11), *yourselves* (1)

TRAP

am (38), *and* (3738), *as* (697), *at* (1012), *had* (91), *has* (16), *have* (168), *shall* (3), *that* (4475)