Longitudinal evidence for simultaneous bilingual language development with shifting language dominance, and how to explain it

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Abstract
Language abilities change with experience, showing immediate priming and interference that build up to long-term learning. Older language production models neglect smaller-scale change and therefore cannot account for acquisition of one, let alone two, languages. Newer models argue that the same domain-general experience-driven incremental learning that supports initial language acquisition also drives continued re-optimisation throughout the lifespan. For bilingual production, instead of simply prioritizing languages by age of acquisition, the balance between languages should reflect their ongoing use. Data from a cross-sequential study of 139 Spanish-English early-sequential bilingual children show that many become more proficient in English than Spanish around third grade, depending on relative exposure to the two languages. Uniquely, the data also show within-subject changes in how quickly children name pictures in each language, thus documenting in part how incremental learning at the small scale of isolated word production accumulates into changes language proficiency and dominance.

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How do a person’s communicative abilities change with experience? The traditional generative view is that they do not. Most models of adult language production, for instance, aim to characterise the functioning of a mature expert system, where processing occurs without durably modifying its mechanisms (e.g. Dell, 1986; Levelt, Roelofs, & Meyer, 1999; see Roelofs, 2018, for an explicit rejection of persistent modification). Such models reflect a tradition wherein it has long been assumed that children acquire abstract knowledge of a language’s syntax, morphology, and phonology by an early age (e.g. as a collection of rules, e.g. Pinker, 1991), and subsequent language development is limited to collecting new words. The underlying assumption is that acquisition quickly produces stable mastery of the target language. This assumption is perhaps most explicit in Chomsky’s (1965 et passim) framework, which posits a distinction between a speaker’s competence and their performance: even young children are assumed to have knowledge of what their language should be, knowing the rules of their language in the same way that a person might ‘know’ a particular word (competence), so difficulties lie in applying that knowledge to particular tasks (performance). When the same framework is applied to the case of learning multiple languages, a person’s first language takes on a special status: whatever language they learn first should continue to be their strongest language throughout their life, meaning, *inter alia*, that it should always be easier for them to speak their first language than any that they acquired later.

Work in the past three decades, however, has brought increasing focus on language plasticity. Empirical work has inspired models in which experience-driven changes throughout the language system lead to changes in the relative availability of syntactic structures (Chang, Dell, & Bock, 2006), phonotactic combinations (Dell, Reed, Adams, & Meyer, 2000), and words (Oppenheim, Dell, & Schwartz, 2010). Contra the competence-performance approach noted above, the simple idea behind these new models is that the same kind of domain-general experience-driven incremental learning that supports a speaker’s initial acquisition of a language also drives continued acquisition and optimisation throughout the lifespan (i.e. language development never stops; e.g. Seidenberg & MacDonald, 1999). In other words, the persistent priming effects that researchers report from laboratory experiments with adults reflect the same mechanism of change that supports early language acquisition. Applied to the question of bilingual production, this ‘incremental learning’ approach further predicts that, instead of simply prioritizing languages in terms of a discrete age of acquisition, the balance between languages should be sensitive to changes in how much a person uses them.

Indeed, extensive work on second language acquisition and education indicates that a person’s ability in a second learned language (L2), if spoken by the majority of their society, often surpasses that in their first (L1; see Oller, Jarmulowicz, Pearson, & Cobo-Lewis, 2011, for a recent review). For instance, as a native
Spanish-speaking child enters an environment where their L2, English, is used more often, their second language can become their best, preferred, most fluent, or 'dominant' language in some if not all contexts (e.g. Dunn & Fox Tree, 2009; Flege, MacKay, & Piske, 2002; Jia & Aaronson, 2003). Although researchers have defined language dominance in many ways, including bilinguals’ estimates of their own competence (Kohnert, Bates, Hernandez, & Diego, 1999), measures of processing speed (i.e. performance) arguably offer better predictive validity (Flege et al., 2002), and, consistent with the idea of speakers learning each time they use a language, dominance changes are manifest not only in children eventually knowing more words in their second language than their first, but also being able to access well-known words more easily (e.g. *dog* outpacing *perro*; Kohnert et al., 1999; Lambert, 1955; Mägiste, 1979).

Most reports of shifts in language dominance come from cross-sectional data (e.g. Kohnert et al., 1999), which is to say testing different children at different ages, assuming that each individual will follow the trajectory suggested by the combined data from the whole group. While it is often risky to draw individual conclusions from group data, logic requires only the highly plausible assumption that an L2-dominant sequential bilingual was L1-dominant early in L2 acquisition. Also, consistent with the idea that dominance shifts could occur within individuals, longitudinal studies of minority language speakers often suggest greater increases in L2 vocabulary than in L1 over time (e.g. Hoff & Ribot, 2017). Consistent with incremental learning accounts, language exposure estimates have been associated with the trajectories of several measures in each language, including performance-based estimates of auditory word recognition (Hurtado, Grüter, Marchman, & Fernald, 2014), but to our knowledge any associations with word production have been limited to competence-based estimates of vocabulary size (Hoff et al., 2012; Winsler, Diaz, Espinosa, & Rodriguez, 1999).

To close this 'performance' gap, we report a within-subjects longitudinal approach to assessing shifts in language dominance, using timed picture naming to track changes in how quickly and accurately 139 native and early-sequential Spanish-English bilingual children retrieve familiar words in each of their languages. As part of a larger study to evaluate a screening test for language impairment, we tested each child in each language, once per year, for up to four years, thus allowing us to assess, for the first time, within-subjects changes in language dominance via a performance-based measure of word production. We tracked vocabulary growth via an untimed test, approximating previous competence-based approaches. Structured interviews of parents and teachers also provide estimates of each child’s language exposure, at home and school, respectively, further allowing us to correlate changes in language accessibility and dominance with changes in language exposure. If changes in language dominance are driven by an individual’s experience, then estimates of their experience should predict at least some of the variation in language dominance trajectories. Because these children all grew up in Spanish-dominant households and subsequently began attending monolingual English or bilingual primary schools, a mismatch between a child’s home and school linguistic environments indexes a major longitudinal
change in their linguistic experience. We therefore show not only that individual children change in their relative abilities in the two languages, but also assess whether their rate of change correlates with this discrepancy between their language environments.

Finally, it is important to note that much of focus on shifts in language dominance has been inspired by concern that learning or improving a second language could reduce a child’s ability to use their first language (Köpke & Schmid, 2004), or that maintaining a first language could impair second language learning. Picture naming speed and accuracy are recognized as a particularly useful means of assessing such difficulties (Schmid & Köpke, 2009). Theoretically, competitive unlearning, on-line competition between language representations, or both could mean that any improvements in L2 access necessarily come at the cost of reducing access to one’s L1 and attempts to maintain L1 could slow acquisition of L2. Although this study cannot address the theoretical question regarding mechanisms of language ‘competition’, it will allow us to address the question of absolute harm.

Method

Overview

We collected these data as part of a larger project aimed at characterizing and screening for possible language impairments among bilingual children; other aspects of this project will be reported separately. Methods for the core experimental procedures are described in this document; methods for supplementary assessments can be found in the Supplementary Information, available online, or in separate articles where referenced as such.

Blocked cyclic picture naming is a widely used paradigm with adults (e.g. Damian, Vigliocco, & Levelt, 2001; Fink, Oppenheim, & Goldrick, 2018; Schnur, Schwartz, Brecher, & Hodgson, 2006) and has been used with children several times before (Charest, 2017; Ladányi & Lukács, 2016). Researchers use speed and accuracy to assess semantic context effects as participants simply try to name pictures that are repeatedly serially presented in related or unrelated blocks. When focusing on context effects, researchers often discard data from the first cycle of each block or treat it as a baseline where the context has not yet been established. At present, however, we simply use blocked cyclic naming as an example of a timed picture naming task, focusing specifically on the first cycle of each block to assess the baseline accessibility of picture names within each language.
Participants

For the larger study, over 300 children from the Austin Texas metropolitan area were recruited via their schools, from Kindergarten and Grades 2 and 4, and tested once per year for up to four years. Because the present report is concerned with shifts in language dominance over time, we include only the 139 Spanish/English bilingual children (74 female, 65 male; see Table 1)) for whom we had two or more years of data and whose parents initially reported more than 50% Spanish use at home (see Home Language Use questionnaire below). Thus, all grew up in Spanish-dominant homes and most also included some amount of English (mean estimated English proportion: 0.234; 75th/25th quartiles: .350/.106); no other languages were spoken at home. All reported normal or corrected-to-normal vision and hearing. Although we have not excluded data from children with suspected language impairments, all claimed results remain if doing so.

<table>
<thead>
<tr>
<th>Test grade</th>
<th>Total</th>
<th>Year of first test</th>
<th>L2 proportion</th>
</tr>
</thead>
<tbody>
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<td></td>
<td></td>
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</tr>
<tr>
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<td>88</td>
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<td>4th</td>
<td>56</td>
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</tr>
<tr>
<td>5th</td>
<td>43</td>
<td>0</td>
<td>23</td>
</tr>
<tr>
<td>Total</td>
<td>139</td>
<td>59</td>
<td>60</td>
</tr>
</tbody>
</table>

Table 1: Participant counts and descriptions by grade. 1 In the first year of the study, some participants completed a different form of the blocked cyclic naming task, so their first-year data are not directly comparable. 2 Three datasets were excluded due to equipment errors.

Language environment questionnaires

Home language use This questionnaire, included in the Supplementary Information, was orally administered each year, via telephone, to each child’s parent or guardian, in whichever language they preferred. It included summary estimates of specific language abilities (Inventory to Assess Language Knowledge; Peña, Gutierrez-Clellen, Iglesias, Goldstein, & Bedore, 2018) and age of first exposure, and hourly estimates of the child’s use of each language during out-of-school hours (Bilingual Input Output Survey; Peña et al., 2018). The respondent sometimes differed from year to year.
School language use  This written questionnaire, included in the Supplementary Information, was completed each year by each child’s schoolteacher. The respondent therefore typically differed from year to year. It included half-hourly estimates of the use of each language, typically assessed for the class as a whole.

Analyses  Although these questionnaires potentially provide a much richer dataset, for this analysis we simply use their first-year home questionnaire to estimate each child’s mean proportion of English versus Spanish use during out-of-school hours, and likewise use the first school questionnaire to estimate their mean proportion of English versus Spanish use during in-school hours (these selections were made a priori, on theoretical bases, without considering other possible associations). We then subtracted the home estimate from the school estimate to estimate the discrepancy between each child’s language environments.

Blocked cyclic picture naming

Materials.  Stimuli were 40 color images, depicting concrete nouns (e.g. dog, pizza), purchased from stock photo libraries. To minimize cross-language transfer, we used different stimuli in each language—selected for high within-language name agreement among even the youngest children—and minimized the use of Spanish-English cognates. The 20 for each language consisted of 4 exemplars from each of 4 semantic categories, plus 4 unrelated objects. Our preliminary norms with bilingual Spanish-English bilingual kindergartners showed that each picture had at least 85% name agreement in the tested language.

Design  The naming task followed a standard blocked cyclic naming design. After an initial familiarization phase, where each picture appeared once in a random order, each picture appeared in one four-item block where every object was an exemplar of the same semantic category (e.g. dog, snake, penguin, elephant), and one four-item block where every object was an exemplar of a different semantic category (e.g. dog, sock, apple, eggs). Within each block, each set of items was presented six times in a pseudo-random order. The order of blocks was itself blocked by condition, and counterbalanced across participants, using 32 stimulus lists that further counterbalanced the order of semantic categories and items within each block. To maximize sensitivity to within-participant changes, each child received the same stimulus list each year. To minimize cross-language transfer, order effects, and language switch costs, each language was tested in a separate session that was administered entirely in the target language. Thus, no language mixing occurred within any part of an experiment session. These sessions also occurred on different days whenever possible (93.1% of cases), with their order counterbalanced across children (testing Spanish first in 49.4% of different-day sessions).
Within-session Procedure  Except for 51 (7%) sessions carried out within a lab setting, all children were tested in the quietest, most isolated location that their school could provide, using a laptop computer for stimulus presentation and response recording.

Familiarisation. At the start of each testing session, children saw each picture once and heard its auditorily presented name, which they then repeated aloud. They were instructed to use these names for the rest of the experiment.

Testing. During testing, children were instructed to name each picture as quickly and accurately as possible as soon as it appeared. Each trial began with a central fixation point displayed for 500ms, then a 250ms blank screen, and finally a single stimulus image in the center of the screen. The stimulus remained for 3500ms or until 1200 ms after the amplitude based voicekey detected a response. A red frame flashed around the stimulus picture when a response was detected to allow the experimenter to monitor voicekey sensitivity. The next trial began 750ms later. If no response triggered the voicekey, a message appeared instructing the experimenter to adjust the microphone or for the participant to speak up. The testing phase for each language consisted of 10x24-trial blocks, lasting approximately 20 minutes.

Apparatus  A 13” MacBookAir laptop computer controlled by MatLab 2010a with PsychToolbox extensions (Brainard, 1992; Pelli, 1997) presented all stimuli and digitally recorded responses via a SteelSeries Siberia V2 Full-Size Gaming Headset.

Response coding  Naming latencies were calculated online, and confirmed offline. Verbal responses were transcribed offline by speakers of the relevant languages. For the purpose of these analyses, responses were coded as correct if they were produced without audible hesitation or correction and differed by no more than one phoneme from the target response from the familiarization procedure or a synonym (bici for bicicleta).

Expressive One-Word Picture Vocabulary Test – Third Edition (EOWPVT-3)

The EOWPVT-3 for English (Brownell, 2000) and the Spanish-Bilingual Edition (Brownell, 2001) are norm-referenced tests of expressive vocabulary. The former is a 170-item picture naming test and the latter includes a subset of the same items. Unlike the blocked cyclic naming test, both are untimed, concerned only with a child’s unfamiliarized accuracy. It is thus a test of vocabulary size rather than accessibility, and somewhat closer to the linguistic definition of ‘competence’. Items are ordered based on difficulty in English.
**Procedure**  We administered the EOWPVT following the standard protocols, but testing continued for 14 items beyond the standard ceiling of six sequential incorrect responses.

**Analyses**  Accuracy was assessed via standard metrics, classifying responses as correct if a child produced an acceptable one-word description of the object in the target language, or a standard phonological variant thereof. Thus, *doggy* would be an acceptable variant of *dog*, but *cappertillar* would not be an acceptable variant of *caterpillar*. We consider raw naming scores, to the standard ceiling, as indicators of absolute vocabulary size.

**General analytical approach**

All analyses use forms of mixed effects regression with maximal random effects for subjects and items, via the lme4 package in R. Error analyses apply logistic regression after excluding equipment errors and non-responses. Naming latency analyses use linear regression of inverse-transformed response times (\(-10000 \times \frac{1}{RT}\)), for correctly detected correct responses only. P values are estimated via the Wald approximation method.

**Results**

**Characterizing the population**

We used parent and teacher interviews to characterize the changes in these initially Spanish-dominant children’s language environments over time (Table 1). According to the parent-based estimates, children’s rate of home English use increased from a mean of .26 in Kindergarten to .41 in 5th grade. In the same timeframe, teachers reported an increase in the classroom school use of English, from a mean of .29 in Kindergarten to .80 in 5th grade, with a particular increase in 3rd grade implementing a practice of transitioning to English dominance in classrooms by that stage. Figure 1 illustrates the distributions of participants’ language use at home and school in their first year of participation. Importantly, although these proportions are positively correlated (\(R = .20, p = .019\))—consistent with an educational approach of instructing students in languages that they actually speak—they are far from perfectly correlated. Their mismatch provides the opportunity to predict changes in language accessibility from changes in language experience, via the plausible assumption that a child’s maturation involves a transition from home-like linguistic environment to one where the school-like environment represents approximately half of their language exposure.
Figure 1: According to parent and teacher estimates, participants' use of English, relative to Spanish, increased over the course of the study, both at home (a) and at school (b). For individual participants, however, there was relatively little correspondence between the initial point estimates for the two settings (c). In Panels (a) and (b), each light grey line represents the estimated trajectory for one participant; the black line represents the estimated trajectory for the group as a whole.

Within-child changes in speed and accuracy in timed picture naming

For these 139 children, we have 366 blocked cyclic picture naming sessions in Spanish and 359 in English. As discussed earlier, we consider only the first naming cycle within each block, both to avoid the semantic effects that are usually the focus of the paradigm and because stronger repetition priming for weaker items (e.g. Griffin & Bock, 1998; Oppenheim, 2018) would attenuate differences that stem from baseline accessibility. This provides 27,588 observations for the mixed effects logistic regression of error data, and 24,889 correct responses for the mixed effects linear regression of naming latencies, illustrated in Figure 2a and b respectively.

To assess experience-driven changes in language accessibility we fit the same mixed effects regression model to both the error rates and the naming latencies, including by-subjects and by-items maximal random effects and the following three fixed effects and their interactions:

1. **Age**, denoting the child’s age in years at the time of testing. To ease interpretation, this continuous variable is centered around 6 years old, the mean age of our youngest cohort.

2. **Language**, a binary-coded contrast where the child’s L1, Spanish, is coded as 0 and their L2, English, is coded as 1.
Figure 2: Changes in language dominance as a function of age, as assessed by (a) error rates and (b) naming latencies in the blocked cycling naming task, and (c) a vocabulary score from the EOWPVT. In the second row (d,e,f), each light grey line represents the cross-language trajectory for one participant; the black line represents that for the group as a whole.
3. **Home-to-school difference in L2 use**, a continuous variable that is centered around its approximate midpoint, .15. As illustrated in Figure 1c, this is the difference between the teacher-estimated rate of English use in their classroom and the parent-estimated rate of English use at home. With this coding, the intercept for a fitted model provides its subject- and item-variability-adjusted estimate for a 6-year-old child naming in Spanish, and main effects and interactions represent deviations from that baseline. Results of these analyses are given in Table 2.
<table>
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<tr>
<th>Errors</th>
<th>Naming latencies</th>
<th>Vocabulary test</th>
</tr>
</thead>
<tbody>
<tr>
<td>27588 obs: 139 subjects, 40 items</td>
<td>24889 obs: 139 subjects, 40 items</td>
<td>707 obs: 138 subjects</td>
</tr>
<tr>
<td>$\beta$</td>
<td>SE</td>
<td>t</td>
</tr>
<tr>
<td>Intercept</td>
<td>-2.083</td>
<td>0.171</td>
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<tr>
<td>Language</td>
<td>0.660</td>
<td>0.240</td>
</tr>
<tr>
<td>Age</td>
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<tr>
<td>Home-school diff</td>
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<tr>
<td>Lang X Diff</td>
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</tr>
<tr>
<td>Age X Diff</td>
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<td>0.179</td>
</tr>
<tr>
<td>Lang X Age X Diff</td>
<td>-0.414</td>
<td>0.231</td>
</tr>
</tbody>
</table>

Table 2: Regression results for the blocked cyclic naming experiment and EOWPVT. Intercept: Language = L1 (Spanish), Age = 6 years old, Home-to-school difference in L2 use = 0.15. Language: L2 (English). Age (in years over 6). Home-to-school difference in L2 use (centered around 0.15)
The youngest children were approximately ten percent slower (Language: $\beta=0.918$, $p=.001$) and twice as likely to err ($\beta=0.660$, $p=.006$) when naming pictures in English compared to Spanish, thus confirming their initial questionnaire-based characterization as Spanish-dominant. Their Spanish naming also significantly improved with age, growing faster (Age: $\beta=-0.661$, $p<.001$) and more accurate ($\beta=-0.188$, $p<.001$). Their English, however, improved more quickly, both in terms of speed (Language X Age: $\beta=-0.339$, $p<.001$) and accuracy ($\beta=-0.345$, $p<.001$). These model parameters thus estimate average shift from L1 dominance to L2 dominance around $6+0.918/0.339 = 8.71$ years if assessed in terms of speed, or around $6+0.660/0.345 = 7.91$ years if assessed in terms of accuracy. This not only replicates the crossover in language dominance reported in previous cross-sectional studies, but further demonstrates that the shift actually occurs within individuals.

A within-subjects longitudinal approach also allows us to examine how individual differences in experience affect language learning. If our observed changes in language dominance are actually driven by changes in a child’s language environment, then the Language X Age interaction should be stronger for those children whose home and school environments differ more. And we see evidence of this interaction in both naming latencies (Language X Age X Home-to-school difference in L2 use: $\beta=-0.563$, $p=.026$) and error rates (albeit less consistently: $\beta=-0.414$, $p=.073$). Thus, the observed changes in language dominance can be linked to changes in language experience.

Converging evidence from vocabulary tests

Other tasks and measures can provide converging evidence through more traditional means. Instead of accessibility, the Expressive One-Word Picture Vocabulary Test aims to assess lexical competence by using unfamiliarized, untimed, picture naming to estimate the size of a child’s vocabulary in each language. Applying the same mixed effects linear regression models to these scores yields similar within-child results (also listed in Table 2).

The youngest children’s vocabulary estimates were significantly greater in Spanish than English (Language: $\beta=-19.46$, $p=.001$). Their Spanish naming scores also significantly improved with age (Age: $\beta=4.38$, $p<.001$). Their English naming scores, however, improved more quickly (Language X Age: $\beta=7.32$, $p<.001$), thus estimating a change from L1 dominance to L2 dominance around $6+19.46/7.32 = 8.66$ years of age, converging with the estimates from the timed naming task. Finally, this interaction was again stronger for those children whose home and school environments differed more (Language X Age X Home-to-school difference in L2 use: $\beta=4.35$, $p=.040$), thus once again linking the observed changes in language dominance with changes in language experience.
Discussion

According to contemporary understanding of how language development works, there is little question that language abilities should correlate with practice. Rather than merely prioritizing one’s languages by age of acquisition, incremental learning accounts predict that a speaker should benefit from each language experience: retrieving ‘dog’ should strengthen links for retrieving ‘dog’ in the future, retrieving ‘perro’ should strengthen links for retrieving ‘perro’, and as a speaker transitions from an L1-dominated environment to an L2-dominated environment these basic practice effects should accumulate into changes in their language dominance. The question is whether they actually do. In this paper we have empirically demonstrated within-speaker shifts from L1-dominance to L2-dominance, as assessed via both performance and competence measures of word production. We have further shown that these changes correspond to gradient changes in the speakers’ linguistic environments, reinforcing the causal link between ongoing experience and demonstrated language dominance.

To our knowledge, this is the first longitudinal evidence of changes in language dominance, as assessed via performance measures of word production. Given the abundance of previous studies of bilingual language acquisition, it might seem odd that such evidence has not previously been reported. Reasons may include the well-known difficulty in conducting longitudinal research (Oller et al., 2011) the common expedient of testing only one language, emphasis on unspeeded tests to measure proficiency or identify disorders, and the use of language-switching and translation tasks to maximize and study cross-language interference (e.g. Kohnert & Bates, 2002; Kohnert et al., 1999).

One potential concern is that, because we tested children in their schools where English tended to be more dominant than in their homes, our data may reflect temporary context-dependent changes in accessibility rather the grander context-independent changes that we have claimed. Although we cannot fully rebut this concern (and to a large extent we agree that memory retrieval should generally incorporate context as a cue), we can offer a few points. Although schools varied in their amount of Spanish use, all used some and could therefore be considered bilingual contexts. Furthermore, each session was run entirely using a single language, thereby creating a maximally supportive context for that language. Moreover, pre-experiment familiarisation of the pictures, with their names in the target language, should have provided an opportunity to re-establish them, which is conservative with respect to our question and claimed findings. Finally, according to most thinking in the wider memory literature, context is typically thought to support weak memories rather than disrupt access to overlearned memories, such as words that a speaker has used hundreds or thousands of times.

Finally, we note that although we have claimed shifts in language dominance as a whole, we have considered only a small component of language: meaning-driven word production. There is mounting evidence that speed of word retrieval may be the most sensitive or labile measure of language dominance. In contrast,
measures of receptive language speed and accuracy often show smaller or later changes in dominance, or no change at all (see Oller et al., 2011, for review). Such observations are consistent with continuous learning, context-dependent, and non-monolithic views of language processing.

**Attrition versus co-improvement.**

One remarkable feature of our results is that the change from L1 to L2 dominance cooccurred with absolute improvements in L1 abilities. That is, speakers’ speed, accuracy, and vocabulary estimates increased in their first language during the period of study; they simply increased more dramatically in their second language. The considerable research focus on first language attrition and suppression (e.g. Köpke & Schmid, 2004; Linck, Kroll, & Sunderman, 2009; Steinhauser, this volume) can lend the impression that language learning is a zero-sum game, where improvement in a second language necessarily harms the first, so our finding of co-improvement might seem like a surprising and even contradictory result. But in fact, both attrition and co-improvement are neatly explained by computational models of incremental language learning that include some mechanism for forgetting or ‘unlearning’.

For instance, in a recent computational project Oppenheim, Dell, & Schwartz (2010) applied a classic incremental learning algorithm (Rescorla & Wagner’s, 1972, ‘delta rule’, originally proposed to describe operant conditioning in rats) to the task of mapping from meaning to words, simulating word retrieval in the same blocked cyclic picture naming that we used here. For each retrieval, they activated several of the network’s semantic features (‘mammalian’, ‘terrestrial’), and it attempted to map them onto an appropriate word (dog). After each attempt, its error-proportional learning algorithm simply strengthened the links from the activated semantic features to the intended word (dog), and weakened the links from those features to any erroneously activated words (bat, whale). Although the network was not ‘born’ knowing any of the correct mappings, it gradually learned to correctly activate dog when it encountered ‘mammal’ and ‘terrestrial’, whale when it encountered ‘mammal’ and ‘aquatic’, and car when it encountered ‘vehicular’ and ‘terrestrial’. More relevant to the current point, when they ran the same network on a blocked cyclic naming experiment—the same task that we used here, where participants repeatedly name a small set of semantically related or unrelated pictures—the same incremental learning algorithm created both 1.) facilitation when the model tried to name the same picture later (dog... dog), thus mapping from the same semantic features to the same word, and 2.) interference when it tried to retrieve a semantically related name instead (whale... dog), thus mapping some of the same semantic features to a different word. When this model performed a naming task without repetition (bat, whale, dog), the unused words in its lexicon grew less accessible each time it retrieved a word with a similar meaning, so it seemed to be losing access to those words. But when the task introduced repetition (dog,
bat, whale, dog), the resulting facilitation outweighed the interference, thus generating net improvement and illustrating how the same learning process can generate both increases and decreases in performance, depending on the ratios of relevant experience. Although this model was originally proposed to account for within-language effects, like cumulative semantic interference, extending it to multilingual production requires only a few assumptions to explain why dog interferes with gato as well as cat (Runnqvist, Strijkers, Alario, & Costa, 2012), thus explaining cases of language co-improvement as well as those of attrition as differences in the ratio of cats to gatos. Switching entirely to a second language may lead to the gradual loss of one’s first language, but interleaving occasional episodes of first-language use may be sufficient to not only undo the damage but even reverse it.

Similarly, in McCloskey and Cohen’s (1989) seminal investigation of ‘catastrophic interference’ in neural networks, although they did not address language per se, they noted that switching from one set of associations (e.g. L1) to a different, competing set (e.g. L2) only produced wholesale forgetting of the originals if the network was trained on the two sets sequentially (as might occur with international adoptees); if the two sets were instead interleaved (more analogous to our participants’ bilingual environments), their networks successfully acquired both. This pattern may explain why reports of L1 attrition have typically focused on cases of L2 immersion, which better approximate McCloskey and Cohen’s sequential training procedure. Thus a strength of this approach is that it illustrates how such domain-specific effects, can emerge from domain general principles.

Conclusion

Experience-driven plasticity is a core feature of all levels of language use and representation. Not only do we continually modify our representations and procedures within a language, dynamically learning and unlearning even well-established words, but we also modify our access to multiple languages.

This is because the same kinds of domain-general incremental learning processes operate at every level of the system, yielding experience driven changes. More generally, such effects reflect the kind of basic incremental learning processes that underlie even complex behaviors like language production.

References


