



Determiner-noun code-switching in Spanish heritage speakers

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Highlights

1. Two linguistic theories of code-switching were compared using behavioral measures.
2. English-dominant heritage speakers of Spanish were tested.
3. Picture naming tasks that elicited determiner-noun code-switches were used.
4. Reaction time and accuracy data did not align with either tested linguistic theory.
5. Results are explained using the WEAVER++ model of speech production.

For Peer Review

Running head: DETERMINER-NOUN CODESWITCHING

Determiner-noun code-switching in Spanish heritage speakers*

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Abstract

Code-switching is prevalent in bilingual speech, and follows specific syntactic constraints. Several theories have been proposed to explain these constraints, and in this paper we focus on the Minimalist Program and the Matrix Language Frame model. Using a determiner-noun picture naming paradigm, we tested the ability of these theories to explain determiner-noun code-switches in Spanish-English bilinguals. The Minimalist Program predicts that speakers will use the determiner from the gendered language, whereas the Matrix Language Frame model predicts that the determiner will come from the language that dominates the syntactic structure in a code-switched utterance. We observed that the bilinguals had slowest naming times and decreased accuracy in Spanish determiner - English noun conditions (“el dog”), and that adding a Matrix Language did not modulate this pattern. Although our results do not align with either theory, we conclude that they can be explained by the WEAVER++ model of speech production.

Keywords: code-switching, determiner-noun phrases, bilingualism, heritage speakers

Determiner-noun code-switching in Spanish heritage speakers

“Mi padre dijo que el car is not working” is an example of a typical code-switched utterance.

Code-switching is defined as the act of switching between two or more languages within a single utterance. This is a phenomenon quite prevalent in bilingual speech as shown by, for example, Clyne (2003), Grosjean (1982), Myers-Scotton (2002), and Poplack (1980) in a variety of different language pairs, including Spanish and English.

There are several reasons why individuals code-switch. Speakers may code-switch when a given word is not accessible in the base language and what the speaker seeks to express is more readily available in the other language (Grosjean, 1982; Heredia & Altarriba, 2001), or when the interlocutor frequently engages in code-switching (e.g., Kootstra, Van Hell, & Dijkstra, 2010).

Additionally, code-switching in conversation may be used to convey socio-cultural or ideological meaning, the nature of the relationship between the interlocutors (e.g., power, respect, intimacy), and the perspective of the speaker (e.g., Bhatt & Bolonyai, 2011; Myers-Scotton, 1993).

Code-switching is studied from different disciplines (for overviews, see Bullock & Toribio, 2009; Isurin, Winford & De Bot, 2009; Van Hell, Litcofsky, & Ting, in press), and a common finding is that code-switching is not a random mixing of languages, but instead follows predictable patterns and is governed by linguistic structural constraints. A linguistic structure that has been studied in several corpora of natural speech is the determiner-noun code-switch. Pfaff (1979) studied Spanish-English code-switching by taking into account the specific characteristics of each language, and how they interact for each syntactic category. In her account of noun phrases, she finds that code-switched determiner-noun phrases used in conversation are more likely to be produced with a Spanish determiner and an English noun than vice versa. Pfaff also noted that in some instances, grammatical gender is lost for nouns borrowed or switched into

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3 Spanish, and masculine gender becomes the default. Jake, Myers-Scotton, and Gross (2002) also
4 studied Spanish-English code-switching and found that determiner-noun switches were prevalent
5 and, similar to Pfaff (1979), that the vast majority of the switches were comprised of a Spanish
6 determiner and an English noun (see also Myers-Scotton & Jake, in press).
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12 Several theoretical models have been developed to explain structural constraints in code-
13 switching (e.g., Belazi, Rubin & Toribio, 1994; Muysken, 2000; Pfaff, 1979; Poplack, 1980;
14 Woolford, 1983), and in this paper we specifically focus on two influential models: the
15 Minimalist Program (MacSwan, 1999) and the Matrix Language Frame model (Myers-Scotton &
16 Jake, 2000; Myers-Scotton, 2002).
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24 *The Minimalist Program*

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27 The Minimalist Program is a generativist approach that sets forth rules to explain how
28 language operates (Chomsky, 1995). MacSwan extended this theory to the realm of bilingual
29 speech by using Minimalist Program principles to predict plausible code-switches (e.g.,
30 MacSwan, 2005). The theory states that in a code-switched utterance, function words are more
31 likely to come from the language that encodes the largest number of uninterpretable features on
32 that particular item (e.g., Liceras, Spradlin & Fuertes, 2005; MacSwan, 2005); uninterpretable
33 features being the aspects of the word that are purely syntactic. Content words, on the other
34 hand, may come from either language. In code-switched noun phrases, it is predicted that the
35 determiner that encodes more features, such as grammatical gender, will be preferred. Thus, in
36 English-Spanish mixed language noun phrases the determiner will come from the language that
37 has grammatical gender, Spanish. Specifically, “la cow” is predicted to occur more frequently
38 than “the vaca” because the Spanish determiner “la” encodes grammatical gender, and the
39 English determiner “the” does not. In the Minimalist Program, code-switching is the union of
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3 two lexicons (MacSwan, 2005), where lexical items from either language are permitted but must
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5 be subjected to the Minimalist Program's feature-checking hypothesis (see MacSwan, 1999, for a
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7 detailed discussion).
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10 *The Matrix Language Frame model*

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12 In contrast to the syntactically based Minimalist Program, the Matrix Language Frame
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14 model was specifically designed to explain code-switched bilingual speech and accounts for both
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16 syntax and processing (Myers-Scotton, 2002). This model labels the two languages being mixed
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18 as the Matrix Language and the Embedded Language. The model assumes an asymmetry
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20 between the languages such that the Matrix Language dominates the syntactic structure of the
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22 code-switched utterance, into which the Embedded Language is inserted. The Matrix Language
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24 provides the grammatical elements in the utterance, such as determiners, pronouns, and
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26 inflectional morphemes (system morphemes). The Embedded Language consists mainly of
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28 nouns, verbs, and adjectives (content morphemes). In addition to these constraints, code-
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30 switched utterances must also be confirmed to be semantically and syntactically sound. In a
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32 code-switched determiner-noun phrase, the Matrix Language Frame model predicts that the
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34 determiner will come from the Matrix Language. For example, a sentence such as “I see the
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36 vaca” is preferred over “I see la cow.” In fact, the model states that the latter is greatly
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38 dispreferred and findings from Jake et al. (2002) support this, showing that such a construction is
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40 rare to be observed at all in bilingual speech. The Matrix Language Frame model was originally
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42 devised to specifically explain code-switched bilingual speech of balanced bilinguals (“classic
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44 code-switching”; Myers-Scotton, 2006). Myers-Scotton argues, though, that it can be extended to
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46 less balanced bilinguals in “composite code-switching.”
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55 *Comparing and testing the Minimalist Program and the Matrix Language Frame model*

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It is to be noted that although the two models both make predictions about code-switching, they originated from different types of data. Support for MacSwan's Minimalist Program is derived mainly from grammaticality judgments, whereas the Matrix Language Frame model is based on naturally-occurring speech. Additionally, even though the two theories differ in their linguistic explanation of code-switches, they still may come to the same prediction in specific cases, but diverge in their predictions for other cases. In the case of determiner-noun code-switched phrases, if the determiner comes from the gendered language and the gendered language is also the Matrix Language, both theories make the same prediction about the code-switching pattern. For example, in the sentence "Yo fui a *la store* a comprar leche," the mixed nominal phrase "la store" is the preferred code-switch according to both theories. The Minimalist Program predicts this switch from a Spanish determiner to an English noun because the Spanish determiner has grammatical gender, and therefore more uninterpretable features than an English determiner. The Matrix Language Frame model predicts this switch because the determiner 'la' comes from the Matrix Language, Spanish. However, if a language such as English is the Matrix Language, the two theories predict competing outcomes. In the sentence "I went to the store," the Matrix Language Frame model predicts the switch "the tienda" as in "I went to the tienda" whereas the Minimalist Program predicts the switch "la store." The question then arises, which of the theories most accurately reflects what is actually produced in bilingual speech?

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Herring, Deuchar, Parafita Couto and Moro Quintanilla (2010) compared the Minimalist Program and the Matrix Language Frame model, targeting their accuracy in predicting the language of the determiner in code-switched nominal phrases produced by Welsh-English and Spanish-English bilinguals. Corpora consisting of conversations from Welsh-English bilinguals living in Wales, UK, and two groups of Spanish-English bilinguals living in Miami, Florida,

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3 were analyzed for this purpose. Determiner-noun code-switches were found to be prevalent in
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5 every group, with 225 instances in the Welsh-English corpus and 148 instances in the two
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7 Spanish-English corpora. In all cases, the language of the determiner was compared to the
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9 language of the sentence (the finite verb was used as the indicator of the Matrix Language). The
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11 Minimalist Program predicts that the determiner must always come from the gendered language,
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13 Welsh or Spanish in this case. The Matrix Language Frame model predicts that the determiner
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15 should come from the Matrix Language, which could be Welsh, Spanish, or English. The
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17 Minimalist Program was able to make a correct prediction 100% of the time in the Welsh-
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19 English corpus and 92% and 93.75% of the time in the two Spanish-English corpora. In
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21 utterances containing a Matrix Language, the Matrix Language Frame model was able to make
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23 correct prediction 98.73% of the time in the Welsh-English corpus, and 97.78% and 94.74% of
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25 the time in the two Spanish-English corpora.
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32 Since both theories were fairly accurate in their predictions, it is not surprising that there
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34 was no statistically significant difference between the two in terms of success of predictions.
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36 Furthermore, the authors note that the Matrix Language was usually either Welsh or Spanish, so
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38 we have little information about how the two theories compare when English is the Matrix
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40 Language.
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43 *The Present Study*

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45 The present study sought to test the Minimalist Program and the Matrix Language Frame
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47 model using well-attested experimental research techniques, again targeting determiner-noun
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49 code-switches in Spanish-English bilinguals. Admittedly, studying code-switching in a
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51 laboratory setting is not as ecologically valid as analyzing naturalistic code-switched speech.
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53 However, our techniques attempt to approach natural code-switching as closely as possible under
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3 highly controlled experimental conditions that allow for a systematic manipulation of the types
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5 of code-switches under study. Our approach takes advantage of laboratory-based techniques to
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7 measure on-line processes of code-switching that provide novel insights into the production of
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9 code-switched speech. The code-switches that are produced are indeed more naturalistic than the
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11 isolated, single word language switching paradigm that is commonly used in the psycholinguistic
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13 literature (for a review, see Bobb & Wodniecka, 2013). We manipulated the language of the
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15 determiner and the Matrix Language in order to uncover the psychological realization of code-
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17 switching within a noun phrase, and to examine the validity of the prediction of these formal
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19 theories for the production of code-switched noun phrases. Additionally, these manipulations
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21 allowed us to control the direction of the switch, thereby expanding upon previous corpora
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23 analyses. We administered a series of picture naming tasks that elicited mixed nominal phrases
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25 both in isolation and within the context of a Matrix Language. Though neither of the models
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27 make explicit predictions about reaction times or accuracy in such a task, by using the Herring et
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29 al. (2010) study as a guide, we were able to extrapolate the general principles of the Minimalist
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31 Program and the Matrix Language Frame model in order to test them in a psycholinguistic
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33 experiment.
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41 The Minimalist Program predicts that the conditions in which the determiner comes from
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43 the gendered language, Spanish, will be easiest and speakers will be faster in producing Spanish
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45 determiner – English noun nominal constructions compared to English determiner – Spanish
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47 noun nominal constructions. We tested this using a determiner-noun picture naming task and a
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49 sentence context picture naming task. The Matrix Language Frame model predicts that the
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51 conditions in which the determiner comes from the Matrix Language will be preferred and have
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3 the fastest processing times, and this prediction was tested in the sentence context picture naming
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8 Our participants were heritage language speakers of Spanish who were also proficient in
9 English. A heritage speaker is an individual raised speaking a language other than English at
10 home, and who holds some degree of competence in both languages (Valdes, 2000). Heritage
11 speakers are unique in that they are often (though not always) less proficient in their L1, and
12 have different strengths and weaknesses than classroom learners of that language or balanced
13 bilinguals (Polinsky & Kagan, 2007). Moreover, heritage speakers are a crucial group for
14 studying the psychological validity of code-switching theories, as heritage speakers engage in
15 code-switching quite often (Bullock & Toribio, 2009; Kagan & Friedman, 2003), driven by
16 factors such as language proficiency and cultural identity.
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29 **Method**

30 *Participants*

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32 The participants were 21 Spanish-English bilinguals (15 females) with a mean age of
33 20.76 ($SD = 3.67$). They were all English dominant native speakers of Spanish, and all but one
34 were currently undergraduate students at Penn State University. Language dominance and
35 proficiency in both languages were assessed using two measures: self-ratings and a bare picture
36 naming task. On a scale from 1 (not proficient at all) to 10 (highly proficient), the average self-
37 rating of Spanish speaking proficiency was 8.6 ($SD = .91$) and the average self-rating of English
38 speaking proficiency was 9.1 ($SD = .86$). On a bare picture naming task (Task 1; see below for a
39 description), the mean reaction time in English was 1025 ms ($SD = 190$) and the mean reaction
40 time in Spanish was 1198 ms ($SD = 175$); the 173 ms difference was statistically significant,
41 $F(1,19) = 9.06, p = .005$ and $F(1,63) = 18.27, p < .001$.¹ Fifteen participants were born in the
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3 United States, and for the remaining six the mean length of stay in the United States was 12
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5 years ($SD = 5.7$). All but one reported to speak Spanish at home with family members on a
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7 regular basis. Due to these specific characteristics of their language background and use, the
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9 participants can be considered English dominant Spanish heritage speakers (e.g., Polinsky &
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11 Kagan, 2007). In the language history and use questionnaire, they also reported engaging in
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13 code-switching on a daily basis. The participants were compensated at a rate of \$10 per hour for
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15 their involvement in the study, and were paid at the conclusion of each of the two sessions.
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19 *Materials*

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21 For all tasks we selected pictures (line drawings) from the Szekely et al. (2004) database
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23 as the basis for the stimulus materials. Each of the three picture naming tasks included the same
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25 set of 64 stimuli, which were translated into both English and Spanish (see Appendix A for a list
26
27 of stimuli). The Spanish stimuli contained 32 masculine Spanish words and 32 feminine Spanish
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29 words. The mean English word length was 4.7 letters (range = 3-7 letters, $SD = 1.2$) and the
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31 mean Spanish word length was 5.8 letters (range = 4-10 letters, $SD = 1.4$). The mean frequency
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33 of the Spanish words was 1.49 ($SD = 0.59$) (Alameda & Cuetos, 1995), and the mean frequency
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35 of the English words was 1.28 ($SD = 0.61$) (<http://celex.mpi.nl>; Baayen, Piepenbrock & Van
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37 Rijn, 1993). The items were matched on English naming agreement (mean = 0.47, $SD = 0.54$)
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39 and Spanish naming agreement (mean = 0.63, $SD = 0.65$; both p 's > .10). Furthermore,
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41 masculine and feminine Spanish nouns were matched on word length and naming agreement
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43 (both p 's > .10). Perfect naming agreement occurs when there is only one word that can be
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45 reasonably used to describe the drawing. An example of a picture with perfect naming agreement
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47 in both English and Spanish is 'heart (corazón).' We also balanced phoneme onset for the
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49 stimuli, and about 2/3 of the English words and the Spanish words began with a stop (p, t, k, d, g,
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3 b; $p > .10$). For the sentence context picture naming task, 10 lead-in sentences were created in
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5 both English and Spanish and checked by two native speakers (who are both highly fluent
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7 Spanish-English bilinguals). In each language, two of the lead-in sentences were used for the
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9 practice session. The remaining eight were divided evenly over the 64 stimuli, such that each
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11 lead-in sentence was presented eight times per block (see Appendix B for a complete list of the
12
13 lead-in sentences).
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16 17 18 *Procedure*

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20 All tasks were completed on a computer in a quiet, dimly lit room with the investigator
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22 present. The study consisted of two sessions one week apart; picture naming tasks 1 and 2 were
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24 completed in the first session and task 3 was completed in the second session. For all tasks,
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26 written instructions indicating the language(s) to be used appeared on the screen either in English
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28 or in Spanish (counterbalanced across participants), and the experimenter reiterated the
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30 directions for each block in English to ensure complete understanding. In addition to the three
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32 critical picture naming tasks, a language history questionnaire was administered at the end of the
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34 second session.
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38 39 *Task 1. Bare picture naming task*

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41 A bare picture naming task was administered first in order to assess bare picture naming
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43 times and to acquire an objective measure of participants' proficiency in their two languages.
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45 The participants named the 64 pictures aloud by saying the noun. The presentation was blocked
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47 by language and the presentation was counterbalanced, such that half the participants named the
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49 English pictures followed by the Spanish pictures, and the other half named first in Spanish then
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51 in English. Each drawing was selected randomly from the list and was preceded by a fixation
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53 cross for 700 milliseconds. The drawing disappeared when a response was detected or when
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3 5000 milliseconds had elapsed, whichever came first. After 600 milliseconds, the fixation cross
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5 was presented followed by the next picture stimulus. A practice session with 4 drawings was
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7 included at the beginning of each block.
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10 11 12 *Task 2. Determiner-noun picture naming task*

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15 Following the bare picture naming task, a determiner-noun picture naming task was
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17 administered in order to test the predictions of the Minimalist Program. Participants were
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19 presented with four blocks of the original 64 drawings and were asked to name them using both a
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21 determiner and a noun. **Written and verbal instructions explicitly directed participants to use a**
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23 **specific language combination, for example, English determiner and Spanish noun. The**
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25 **instructions always provided an example phrase to the participants to further illustrate the desired**
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27 **response. For Spanish determiner – English noun phrases, the example kept the grammatical**
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29 **gender of the Spanish translation equivalent (e.g., “la table”).** To test code-switching in both
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31 directions as well as non-switched noun phrase production, the determiner and the noun could be
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33 in either English or Spanish, for a total of four conditions: English determiner - English noun
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35 (e.g., “the clock”), English determiner - Spanish noun (e.g., “the reloj”), Spanish determiner -
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37 English noun (e.g., “el clock”), and Spanish determiner - Spanish noun (e.g., “el reloj”). The
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39 stimulus presentation was blocked by condition; participants named a picture in one of the four
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41 conditions, according to on-screen and verbal instructions at the beginning of each block. The
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43 blocks were counterbalanced across participants. As in the bare picture naming task, each
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45 drawing was preceded by a fixation cross for 700 milliseconds, followed by a 600 millisecond
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47 blank screen, and the drawing disappeared when a response was detected or when 5000
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3 milliseconds had elapsed. A practice session with 4 pictures was included at the beginning of
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5 each block.
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8 *Task 3. Sentence context picture naming task*

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10 In the final picture naming task, participants were presented with a lead-in sentence such
11 as “On the next screen there is...” that acted as a matrix language, and then with one of the 64
12 pictures, selected at random. This task tests predictions **made on the basis** of the Minimalist
13 Program, but the addition of the matrix language in this task allowed us to also specifically test
14 the predictions **that follow from** the Matrix Language Frame model. The participant was asked to
15 complete the sentence by naming the picture aloud using both a determiner and a noun, as in the
16 determiner-noun picture naming task (task 2). Each of the four conditions of the determiner-noun
17 picture naming task (task 2) was combined with both a lead-in sentence in English (e.g., “**This is**
18 **a picture of...**”) and a lead-in sentence in Spanish (e.g., “**Este es un dibujo de...**”), for a total of
19 eight conditions. Thus, the eight conditions comprised all permutations of the two languages.
20 These conditions were presented in blocks, counterbalanced across participants, and participants
21 were instructed at the beginning of each block by an on-screen cue (which was reiterated
22 verbally) which languages to use. **For example, in English-Spanish conditions, participants were**
23 **told to name the picture using an English determiner and a Spanish noun.** As in the previous
24 tasks, each drawing appeared on the screen until a response was detected (up to 5000
25 milliseconds), was preceded by a fixation cross for 700 milliseconds, and was followed by a
26 blank screen for 600 milliseconds. As in tasks 1 and 2, a practice session with four trials was
27 included at the beginning of each block.
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52 *Data cleaning, scoring, and analysis*

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3 The data from one participant were excluded entirely, because of extremely low
4 proficiency in Spanish (less than .17 accuracy on the bare picture naming task), bringing the total
5 to 20 participants. For each response, we determined the accuracy and reviewed the reaction
6 times. English responses were considered accurate when they matched the label given to the
7 picture in the Szekely et al. (2004) database and Spanish responses were considered accurate
8 when they matched the translation equivalent, which was determined by two native Spanish
9 speakers. This strategy was chosen to cover potential variations in the vocabulary of our
10 bilinguals. It was not necessary to make changes to the English stimuli, since the labels in the
11 Szekely et al. database are standard American English. Reaction times associated with inaccurate
12 responses (23%, 20%, and 20% for tasks 1, 2 and 3 respectively), voice key registration errors
13 (3%, 2%, and 1% for tasks 1, 2 and 3 respectively), responses below 300 milliseconds or above
14 3000 milliseconds (2%, 1%, and 2% for tasks 1, 2, and 3 respectively), or above 2.5 standard
15 deviations of a participant's mean (2%, 2%, and 2% for tasks 1, 2, and 3 respectively) were
16 removed from the reaction times analyses. The accuracy data included incorrect responses and
17 omissions (i.e., no response was given within the 5000 ms response window). For the bare
18 picture naming task (task 1), we analyzed the difference between the English and Spanish
19 reaction times and accuracy using one-way ANOVAs, by participants ($F1$) and by items ($F2$).
20 For the determiner-noun picture naming task (task 2), 2 (determiner language: English, Spanish)
21 by 2 (noun language: English, Spanish) ANOVAs were performed by participants ($F1$) and by
22 items ($F2$), separately so for the RT and the accuracy data. For the sentence context and
23 determiner-noun picture naming task (task 3), 2 (sentence context language: English, Spanish),
24 by 2 (determiner language: English, Spanish) by 2 (noun language: English, Spanish) ANOVAs
25 were performed by participants ($F1$) and by items ($F2$), separately so for the reaction time and
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3 the accuracy data. In these ANOVAs, the factors determiner language, noun language, and
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5 sentence context language were treated as within-participants factors in the by-participants
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7 analysis, and as between-item factors in the by-items analysis.
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10 **Results**

11 *Bare picture naming task*

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15 In the bare picture naming task, the mean naming time for English nouns was 173 ms faster than
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17 for Spanish nouns, $F1(1,19) = 9.06, p = .005$ and $F2(1,126) = 18.27, p < .001$ ($M = 1025$ ms, SD
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19 = 190 ms for English nouns; $M = 1198$ ms, $SD = 175$ ms for Spanish nouns). Likewise, picture
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21 naming accuracy was higher for the English than for the Spanish nouns, $F1(1,19) = 116.23, p <$
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23 $.001$ and $F2(1,126) = 47.60, p < .001$ ($M = .84, SD = .06$ for English nouns; $M = .59, SD = .09$
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25 for Spanish nouns). Participants were both faster and more accurate in bare picture naming in
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27 English than in Spanish.
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31 *Determiner-noun picture naming task*

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34 In the determiner-noun picture naming task, participants were asked to name pictures
35
36 using both a determiner and a noun, in either code-switched or not code-switched conditions.
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38 Mean reaction times and accuracy are depicted in Figures 1 and 2, respectively. In the reaction
39
40 times analysis, the main effect of determiner language was significant, $F1(1, 19) = 63.65, p <$
41
42 $.001$ and $F2(1, 252) = 821.56, p < .001$, as was the main effect of noun language, $F1(1, 19) =$
43
44 $19.37, p < .001$ and $F2(1, 252) = 36.14, p < .001$. Importantly, the interaction between
45
46 determiner language and noun language was also significant, $F1(1, 19) = 95.57, p < .001$ and
47
48 $F2(1, 252) = 573.58, p < .001$. Simple effects analyses showed that in comparing the two code-
49
50 switched conditions, the Spanish determiner - English noun condition was substantially (i.e., 577
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52 ms) slower than the English determiner - Spanish noun condition, $t1(19) = 6.26, p < .001$ and
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3 $t_2(126) = 20.08, p < .001$. Additionally, production of noun phrases entirely in Spanish was 202
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5 ms slower than production of noun phrases in English, $t_1(19) = 4.33, p < .001$ and $t_2(126) =$
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7
8 $9.43, p < .001$.

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10 <Insert Figures 1 & 2 about here>

11
12 ANOVAs on the accuracy data yielded a significant main effect of determiner language,
13
14 $F_1(1, 19) = 20.75, p < .001$ and $F_2(1, 252) = 26.44, p < .001$, and of noun language, $F_1(1,19) =$
15
16 $53.41, p < .001$ and $F_2(1, 252) = 22.58, p < .001$.² The interaction between determiner language
17
18 and noun language was also significant, $F_1(1, 19) = 40.48, p < .001$ and $F_2(1, 252) = 36.09, p <$
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20 $.001$. Simple effects analyses on the accuracy data yielded a significant difference between the
21
22 two code-switched conditions, such that the Spanish determiner – English noun condition was
23
24 more accurate than the English determiner – Spanish noun condition, $t_1(19) = 3.54, p < .001$ and
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26 $t_2(126) = 3.43, p < .001$. A comparison of the single language non-switched conditions showed
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28 that producing English noun phrases was more accurate than producing Spanish noun phrases,
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30 $t_1(19) = 9.32, p < .001$ and $t_2(126) = 5.69, p < .001$.

31 32 33 34 35 36 37 *Sentence context picture naming task*

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39 In the sentence picture naming task, participants read lead-in sentences prior to naming
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41 the pictures using a determiner and a noun. Mean reaction times and accuracy are depicted in
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43 Figures 3 and 4, respectively. The reaction time analysis yielded significant main effects of
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45 determiner language, $F_1(1, 19) = 88.43, p < .001$ and $F_2(1, 504) = 718.25, p < .001$, and noun
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47 language, $F_1(1, 19) = 32.72, p < .001$ and $F_2(1, 504) = 81.69, p < .001$. As in Task 2, the
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49 interaction between determiner language and noun language was also significant, $F_1(1,19) =$
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51 $55.24, p < .001$ and $F_2(1, 504) = 656.712, p < .001$. Simple effects analyses comparing critical
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60 contrasts showed that in the code-switched conditions, the Spanish determiner – English noun

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3 condition was slower than the English determiner - Spanish noun condition, both when paired
4 with an English sentence, $t1(19) = 6.36, p < .001$ and $t2(126) = 19.08, p < .001$, and with a
5 Spanish sentence, $t1(19) = 4.71, p < .001$ and $t2(126) = 14.89, p < .001$. No significant main
6 effect of sentence language was found, and the 3-way interaction between the three factors was
7 not significant either.
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13 <Insert Figures 3 & 4 about here>
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17 The ANOVAs on the accuracy data showed main effects of determiner language, $F1(1,$
18 $19) = 37.45, p < .001$ and $F2(1, 63) = 49.59, p < .001$, and of noun language, $F1(1, 19) = 32.47,$
19 $p < .001$ and $F2(1, 504) = 18.25, p < .001$. The interaction between determiner language and
20 noun language was also significant, $F1(1, 19) = 18.38, p < .001$ and $F2(1, 504) = 40.65, p < .001,$
21 and the interaction between sentence language and determiner language reached significance in
22 the by-subjects analysis, $F1(1, 19) = 7.90, p = .013$, but not in the by-items analysis. Again, no
23 significant main effect of sentence language was found, or a 3-way interaction between the three
24 variables. Simple effects analyses of the accuracy data showed that the Spanish determiner –
25 English noun code-switched condition was more accurate than the reverse, which was marginally
26 significant in the by-participants analysis when preceded by an English sentence context, $t1(19)$
27 $= 1.73, p = .093$, significant in the by-items analysis when preceded by an English sentence
28 context, $t2(126) = 2.73, p = .007$, and significant when preceded by a Spanish sentence context,
29 $t1(19) = 2.53, p = .005$ and $t2(126) = 2.94, p = .004$.
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48 Discussion

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50 Using Spanish-English bilinguals who were all heritage speakers, we experimentally
51 tested the predictions **that follow from** the Minimalist Program and the Matrix Language Frame
52 model with regard to language mixed nominal constructions. Whereas previous work on this
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3 topic has focused predominantly on corpus data, we approached the question from a new angle
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5 by systematically manipulating the language of the determiner, of the noun, and of the sentence
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7 context, and analyzed reaction times and accuracy data, in order to gain insight into cognitive
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9 mechanisms of the production of language-mixed nominal constructions. These types of code-
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11 switched phrases were elicited in the determiner-noun picture naming task and the sentence
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13 context picture naming task. We found that the English-dominant heritage speakers were slower
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15 in producing a code-switched determiner noun phrase that contained a Spanish determiner and an
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17 English noun than vice versa. Furthermore, we found that the Matrix Language context did not
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19 modulate speed or accuracy of the production of code-switched noun phrases: the results from
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21 the sentence context picture naming task mirrored those of the determiner-noun picture naming
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23 task.
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29 To examine to what extent our findings were consistent across the different variants of
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31 the picture naming task, and were not driven by specific task-demands, we performed a 2
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33 (Determiner Language: English, Spanish) by 2 (Noun Language: English, Spanish) by 2 (Task:
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35 Task 2, Task 3) ANOVA, separately so for the reaction times and accuracy data, and found no
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37 main effect of Task, or any interactions with the factor Task in either analysis. The observed
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39 pattern is, therefore, not different across the different task types. In both picture naming tasks
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41 (with both sentence context languages), the Spanish determiner - English noun code-switched
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43 noun phrases yielded the slowest reaction times and lower accuracy as compared to the English
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45 determiner - English noun condition.
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50 As mentioned earlier in the paper, corpus studies have observed that Spanish-English
51
52 bilinguals have a tendency to default to the masculine determiner “el” when code-switching from
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54 Spanish to English within a noun phrase, regardless of the grammatical gender of the translation
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3 equivalent of the English noun (Jake et al., 2002; Pfaff, 1979; see also Valdes Kroff, 2012).
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5 According to this observation, it would be natural for a speaker to, for example, say “el store,”
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7 despite the fact that store (*tienda*) in Spanish is feminine and takes the feminine determiner “la.”
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9 To examine whether this tendency was also present in our group of bilinguals and was
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11 responsible for the results we observed, we examined the specific types of errors being produced,
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13 focusing on instances in which the incorrect gender was produced. Our analysis of the errors
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15 made in the Spanish determiner - English noun nominal constructions showed that our
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17 participants did not rely on this strategy. For feminine nouns in the determiner-noun picture
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19 naming task, participants defaulted to the masculine determiner in the Spanish determiner -
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21 English noun condition only 7.2% of the time. In the sentence context picture naming task the
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23 results were the same, with participants using the masculine determiner “el” with 7.2% of
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25 feminine nouns. None of our participants exclusively produced masculine determiners. This is in
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27 line with Liceras et al. (2008), who argue that the tendency for bilingual speakers of Spanish and
28
29 English to default to the masculine gender is only present in English-Spanish bilinguals, not in
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31 early Spanish-English bilinguals (i.e., the type of bilinguals who participated in our study).
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33 Additionally, evidence from Italian-German early bilinguals shows that when switching
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35 languages between a determiner and a noun, the gender of the noun is preserved in the utterance
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37 (Cantone & Müller, 2008).
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46 Our finding that Spanish determiner – English noun phrases are difficult to produce was
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48 not predicted by either of the tested theories. Because the Minimalist Program argues that in a
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50 code-switched noun phrase, the determiner will come from the gendered language, **the corollary**
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52 **prediction regarding** reaction times is that Spanish determiner - English noun code-switched
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54 phrases should be produced faster than English determiner - Spanish noun code-switched
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3 phrases, which is the opposite of what we found in both task 2 and task 3. The Matrix Language
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5 Frame model, which argues that in a code-switched noun phrase the determiner will come from
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7 the Matrix Language, was not supported due to the lack of effect of sentence context language on
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9 reaction times or accuracy in task 3.
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13 One possible explanation for the slower naming times and lower accuracy in the Spanish
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15 determiner - English noun condition is that the English noun is harder to retrieve than the
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17 Spanish noun. However, both latency and accuracy data of the bare noun picture naming tasks
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19 show participants were significantly faster and more accurate in naming English nouns than
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21 Spanish nouns. This indicates that the increased reaction times and decreased accuracy in the
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23 Spanish determiner - English noun condition comes from difficulty selecting the Spanish
24
25 determiner, rather than from difficulty retrieving the English noun.
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29 An alternative explanation is that determiner-noun code-switches are rare, and come
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31 unnatural to the Spanish-English bilinguals we tested. However, it is not the case that these
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33 determiner-noun code-switches are rare; in fact, it is quite the opposite. Determiner-noun
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35 switches are frequently observed in a wide range of bilingual corpora (Herring et al., 2010;
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37 Licerias et al., 2005; Pfaff, 1979; Timm 1975; Woolford, 1983), attesting that bilingual code-
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39 switchers are likely to use this construction in everyday speech. Moreover, switching from a
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41 Spanish determiner to an English noun is also not more unnatural than switching from an English
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43 determiner to a Spanish noun. On the contrary: Milian (1996; as cited in Myers-Scotton & Jake,
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45 1997) and Licerias et al. (2005) have shown that the Spanish determiner - English noun switch is
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47 more common than a switch in the opposite direction.
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53 Although our findings do not support either the Minimalist Program or the Matrix
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55 Language Frame model, this does not imply that the models are incorrect.⁴ What our results do
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3 suggest, however, is that these models need further adaptation in order to account for the online
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5 processes of code-switching. One or both of the models may be effective in explaining the
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7 frequency of particular code-switches in naturalistic speech (Herring et al., 2010), but not the
8
9 ease of the actual production of code-switched determiner-noun phrases. An interesting avenue
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11 for further research would be to investigate this disparity between what is easiest to produce and
12
13 what is most frequently produced, and to also relate this to the comprehension of code-switched
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15 determiner-noun phrases. For example, studying the comprehension of visually presented code-
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17 switched nominal phrases, Dussias (2001) found reading times to align with data obtained from
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19 bilingual corpora: Reading Spanish determiner – English noun phrases (either singular or plural)
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21 was 14 ms faster than reading a switch in the opposite direction. One way to further understand
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23 the nature of the relation between ease of production and comprehension on the one hand and
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25 frequency of production on the other, is to collect multiple types of data (lab-based and natural
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27 speech studies of codeswitching) from a variety of bilinguals, including habitual and non-
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29 habitual codeswitchers, bilinguals who are functioning in a single or dual language context, or
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31 bilinguals with varied L1 and L2 proficiency levels (cf. Green, 2011; Green & Wei, in press;
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33 Grosjean, 1997). Another factor to take into account are the cognitive implications of processing
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35 externally induced codeswitches, i.e., non-spontaneous language switches where bilinguals
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37 switch languages prompted by an external cue or respond to an externally generated switch, and
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39 internally induced codeswitches, i.e., spontaneously generated language switches (Gullberg,
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41 Indefrey, & Muysken, 2009). Lab-based studies on the production of codeswitches often focus
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43 on externally-induced codeswitches (for exceptions, see Gollan & Ferreira, 2009; Kootstra, van
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45 Hell, & Dijkstra, 2010) as do studies on the comprehension of codeswitches, whereas corpus
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47 studies typically examine internally induced codeswitches. Studying code-switching from formal
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3 linguistic, sociolinguistic, and psycholinguistic perspectives, using different research techniques,
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5 will tap into different aspects of codeswitching, and will advance the formation of
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8 comprehensive models of code-switching predictive of multiple aspects of bilingual speech.
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10 The Minimalist Program or the Matrix Language Frame model cannot fully explain the
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12 observed results, nor can the alternative explanations outlined above. In the remainder of the
13
14 discussion, we therefore relate the present findings to current speech production models
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16 (typically based on unilingual speech), and explore whether our findings can be explained by
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18 how these models describe how noun phrases and determiners are retrieved and produced in
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20 speech.
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24 In Roelofs' (1992) WEAVER⁺⁺ model, producing an open class word, such as a noun, is
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26 argued to be a feedforward process flowing from conceptual preparation, to lexical selection, to
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28 morphophonological encoding, to phonetic encoding, all the way to articulation. In the
29
30 conceptual stage, the meaning of the intended word is activated. Once this lexical concept is
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32 made available, the lemma for that concept is retrieved from the mental lexicon. A lemma is an
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34 abstract form of a word that includes basic semantic and syntactic information. For example, a
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36 lemma will indicate whether the to-be-produced word is a verb or a noun. Once the speaker has
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38 retrieved this basic information, he or she can begin building the word form. The first stage is
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40 morphological encoding, which involves retrieving the appropriate morphemes for the word. In
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42 addition to the root of the word, retrieving morphemes indicating gender, number or tense, may
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44 be necessary. Next, the phonemes associated with each morpheme are activated. Before the
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46 speaker can articulate the word, however, he or she must segment the phonological form into the
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48 correct syllables and then finally apply the appropriate phonetic gestures. These gestural scores
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50 will inform the articulatory system how to produce the sounds. At all stages after the
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3 phonological word form is retrieved (including after articulation), self-monitoring occurs. If an
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5 incorrect word form has been selected or produced, the process may begin again at the
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7 conceptual stage. The process for selecting and producing a determiner is different (e.g., Alario,
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9 Ayora, Costa & Melinger, 2008; Levelt, 1993; Miozzo & Caramazza, 1999), and complicated by
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11 the fact that many languages, such as our target language Spanish, encode grammatical gender
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13 information on the determiner. There are different views on how the selection and production of
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15 determiners is achieved. The WEAVER⁺⁺ model proposes that retrieval of grammatical gender
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17 information only occurs when gender agreement is required by the syntactic properties of the
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19 language, such as in the production of a noun phrase in Spanish. The grammatical gender of the
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21 Spanish noun must be retrieved in order to produce the correct determiner, either “el” or “la.”
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23 This gender information is accessed directly after lemma retrieval (Roelofs, 2006), and in a
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25 language with a two-gender system, the two genders compete for selection in production.
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27 Empirical evidence for this competition comes from studies showing this gender (in)congruency
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29 effects in picture-word and picture-picture distractor paradigms (e.g., La Heij, Mak, Sander &
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31 Willeboordse, 1998; Schriefers, 1993). For example, Schriefers (1993) found that naming
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33 latencies for producing a noun phrase in response to a picture were shorter when the distractor
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35 has the same gender as the target. When the target and the distractor differ in grammatical
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37 gender, an inhibitory effect was obtained. This gender congruency effect was not observed in
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39 bare noun production, because, as argued by the authors, gender information is not required for
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41 production of a noun in isolation.
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50 A similar viewpoint is taken by Caramazza’s Independent Network model of speech
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52 production (Caramazza, 1997). The Independent Network model also postulates that
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54 grammatical gender is only retrieved when it is required for agreement. However, in contrast to
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3 WEAVER⁺⁺, the Independent Network model argues that gender selection is a non-competitive
4 process that follows from selecting a lexical item, but that selection of the appropriate determiner
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6 to complete the noun phrase is a competitive process similar to selecting a lexical node. This
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8 implies that selecting a gendered determiner from two or more alternatives should not cause a
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10 processing cost. Caramazza's (1997) explanation of the gender congruency effect reported in
11
12 previous studies is that there is competition at the level of determiner selection, not gender
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14 selection. Evidence for this position comes again from the picture-word distractor paradigm, and
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16 makes use of the Dutch determiner system, in which there are two singular determiners ("de" for
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18 common gender nouns and "het" for neuter gender nouns) and one shared determiner "de" for
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20 plural nouns. Schiller & Caramazza (2003) observed the interference expected in singular target-
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22 distractor pairs with incongruent grammatical gender and two different determiners, but no
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24 interference when the gender incongruent target-distractor pairs shared a determiner.
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32 An opposing view is postulated by Cubelli, Lotto, Paolieri, Girelli and Job (2005), who
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34 found evidence for gender selection even in bare noun production in Italian that has two genders.
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36 In a picture-word interference paradigm, naming latencies for bare nouns were significantly
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38 longer when the distractor word was of the same grammatical gender as the to-be-named picture.
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40 Additionally, an effect of semantic relatedness was found in which a picture was named slower
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42 when the distractor word was related in meaning. Their explanation is that both semantic and
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44 gender information must be accessed before the phonological form, and that these are
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46 competitive processes, since an inhibitory effect was found on congruent trials. The competition
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48 at these two levels must be resolved before the lexical form of a noun is retrieved.
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53 Our results are more in line with the assumptions of the WEAVER⁺⁺ model of speech
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55 production, **as we do not find evidence for the retrieval of grammatical gender when it is not**
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3 required for agreement. We propose that the condition in which bilinguals produce the
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5 determiner in English and the noun in Spanish is comparable to a bare picture naming situation,
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7 because gender information for the noun does not need to be retrieved in order to produce the
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9 English determiner. The WEAVER++ model states that no gender information should be
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11 activated in this specific condition, as the model argues that gender is only retrieved in situations
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13 that require gender retrieval for agreement (e.g., noun phrases). In contrast, the Spanish
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15 determiner - English noun condition requires grammatical gender agreement in order to produce
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17 the correct Spanish determiner, because selection of a Spanish determiner depends on the gender
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19 of the noun. This implies that there is gender information activated in the Spanish determiner -
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21 English noun condition, and the two gender alternatives, masculine and feminine, compete for
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23 selection. This competition effect is reflected in longer reaction times and decreased accuracy in
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25 this specific condition.
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32 A similar model of speech production has been applied to code-switching recently by
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34 Myers-Scotton and Jake (2013), but in relation to the patterns observed in corpus data. The au-
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36 thors argue that after the speaker's semantic and pragmatic intentions are established, content
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38 morphemes (e.g., nouns, verbs, adjectives) and early system morphemes (e.g., determiners, plu-
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40 ral markers) are selected and sent to a formulator, which then selects late system morphemes
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42 (e.g., possessive markers). This means that for mixed nominal phrases in Spanish and English,
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44 the determiner is selected early on in the process, directly after the noun. Myers-Scotton and Jake
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46 further state that since Spanish determiners have the additional inherent feature of grammatical
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48 gender, they are more salient to the speaker than the English determiner 'the' and will always be
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50 preferred, unless English is the Matrix Language. They argue that switching between a Spanish
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52 determiner and an English noun is so prevalent and seemingly effortless because of the two
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aforementioned factors: the salience and early selection of Spanish determiners. Conversely, our data show that the salience of grammatical gender does not make Spanish determiners faster to produce, and in fact the opposite pattern occurs. This salience may be the cause of the prevalence of Spanish determiner-English noun phrases in naturalistic speech, but our results indicate that there is still a cost associated with producing these types of switches, most likely due to competition at the level of determiner selection. **Additionally, the Matrix Language manipulation in the present study did not affect production whatsoever.**

To conclude, the present experimental tests of bilinguals' production of code-switched determiner-noun nominal constructions, measuring the online processing of code-switched noun phrases, did not confirm the **corollary predictions** of two influential linguistic models describing structural constraints that govern the grammaticality of code-switched utterances: the Minimalist Program and the Matrix Language Frame model. The present study is among the first to test the predictions **that follow from** the Minimalist Program and the Matrix Language Frame model in a laboratory setting, measuring online production of code-switched utterances (cf. Bultena, Dijkstra, & Van Hell, 2015a; b; Kootstra, Van Hell, & Dijkstra, 2010; Kootstra, Van Hell, & Dijkstra, 2012; Myers-Scotton, 2006), to examine the psychological validity of these linguistic theories. Among interesting future directions would be a follow-up study testing speakers of two languages that overtly mark gender (e.g., Spanish-Italian or German-French bilinguals), to examine how grammatical gender (and congruent or incongruent gender mapping across languages) affects production times of code-switched determiner-noun phrases in bilingual speakers. **It would also be of interest to test other populations of Spanish-English bilinguals, such as more balanced bilinguals and Spanish dominant bilinguals, to see how language dominance influences the ease with which code-switched noun phrases are produced, and to examine the**

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3 extent to which the present results also hold for these bilingual groups. Indeed, it may be the case
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5 that balanced or Spanish-dominant bilinguals will experience less difficulty producing Spanish
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7 determiner – English noun phrases than our current heritage speaker sample. Additional research,
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9 using a wider range of experimental tasks, is needed to gain more insight into the psychological
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11 realm of linguistics theories and connect these to psycholinguistic models of language
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13 production, but the present study indicates that descriptive corpus analyses of code-switched
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15 nominal constructions do not necessarily align with online measures of the actual production of
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17 such constructions (cf. Van Hell, Litcofsky, & Ting, in press).
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Footnotes

1. The data of one participant had be excluded, see results section for more details.
2. We performed an alternate accuracy calculation for the determiner-noun picture naming task and the sentence context picture naming task, with trials in which the participant did not respond removed. For the determiner-noun picture naming task, the accuracy results (percentage correct) are as follows: English determiner - English noun 94.68%; Spanish determiner - Spanish noun 81.23%; Spanish determiner - English noun 87.57%; English determiner - Spanish noun 79.18%. For the sentence context picture naming task, the results for conditions with an English Matrix Language are as follows: English determiner - English noun 92.14%; Spanish determiner - Spanish noun 81.63%; Spanish determiner - English noun 87.59%; English determiner - Spanish noun 80.84%. The results for conditions with a Spanish Matrix Language are as follows: English determiner - English noun 93.75%; Spanish determiner - Spanish noun 80.83%; Spanish determiner - English noun 86.59%; English determiner - Spanish noun 79.54%. The accuracy is overall higher than the accuracy calculation used for our analyses, but it follows the same pattern for both tasks.
3. Note that both models make predictions concerning the translation equivalent determiner with the correct grammatical gender, such as “la cow.” However, a Spanish-English bilingual may default to the masculine gender when code-switching and produce a phrase such as “el cow” (Jake et al., 2002; Pfaff, 1979;Valdes Kroff, 2012). As explained in more detail in the Discussion section, we will take this into account when analyzing our data and interpreting the findings.
4. It should be noted that the present divergence between corpus data and experimental data on codeswitching is not an isolated observation. In a recent series of lab-based studies, we tested the processing implications of the corpus-based lexical triggering theory (Clyne, 2003), stating that

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3 code-switches are more likely to occur near triggers (cognates, homophones, proper nouns) than
4 non-triggers (see also Broersma & De Bot, 2006). In a confederate-scripted picture description
5 study, it was found that producing cognates indeed increased the likelihood of also producing a
6 codeswitch, but only when the confederate had also produced a codeswitch in the previous trial
7 (Kootstra, 2012; Kootstra, van Hell, & Dijkstra, in revision). However, in a shadowing study
8 (Bultena, Dijkstra, & Van Hell, 2015a) and a self-paced reading study (Bultena, Dijkstra, & Van
9 Hell, 2015b), cognate triggers did not affect the ease of producing or comprehending a
10 codeswitched sentence.
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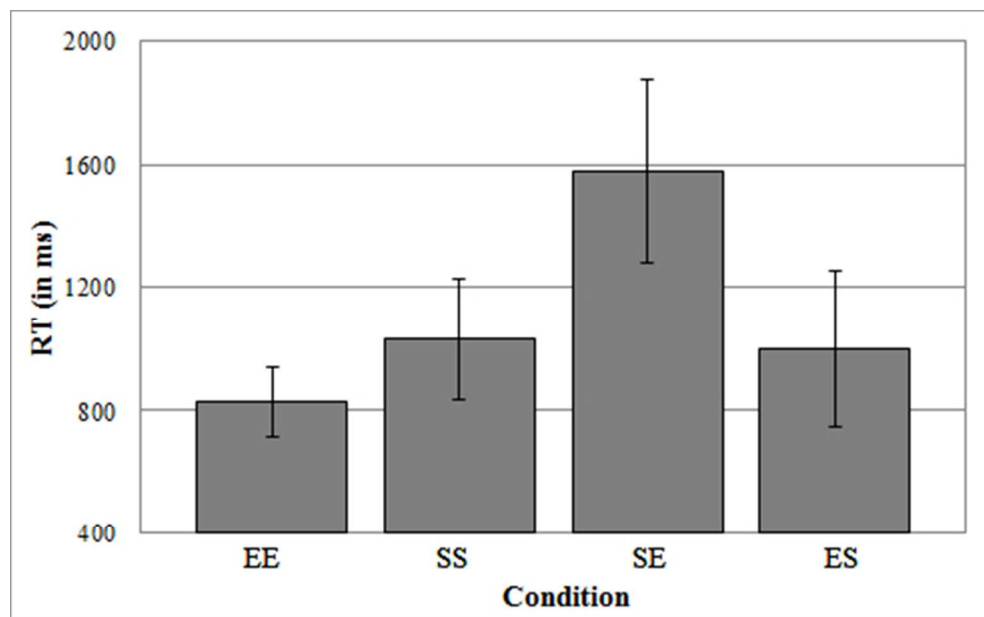


Figure 1. Mean Determiner-Noun Picture Naming Times (Task 2). EE = English determiner, English noun; SS = Spanish determiner, Spanish noun, SE = Spanish determiner, English noun; ES = English determiner, Spanish noun. Error bars depict Standard Deviations.
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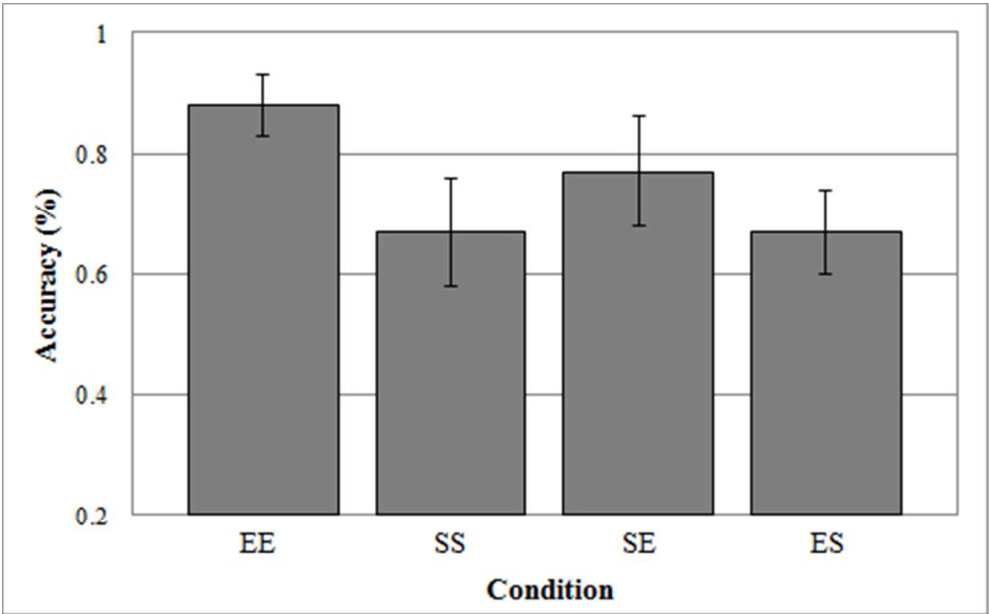


Figure 2. Mean Determiner-Noun Picture Naming Accuracy (Task 2). EE = English determiner, English noun; SS = Spanish determiner, Spanish noun, SE = Spanish determiner, English noun; ES = English determiner, Spanish noun. Error bars depict Standard Deviations.
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Review

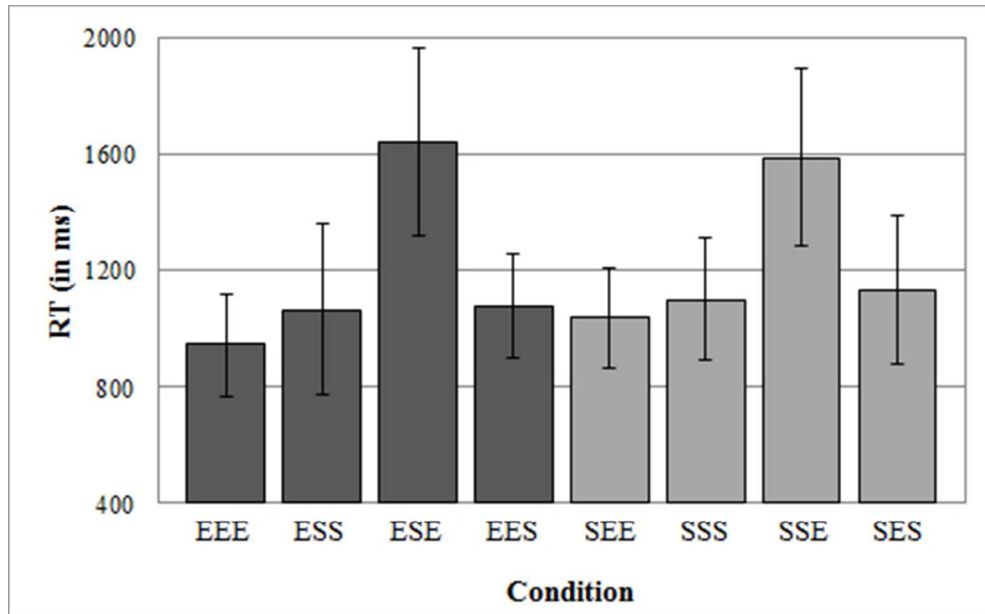


Figure 3. Mean Sentence Context Determiner-Noun Picture Naming Times (Task 3). Dark bars indicate English sentence, light bars indicate Spanish sentence. EEE = English sentence, English determiner, English noun; ESS = English sentence, Spanish determiner, Spanish noun; ESE = English sentence, Spanish determiner, English noun; EES = English sentence, English determiner, Spanish noun; SEE = Spanish sentence, English determiner, English noun; SSS = Spanish sentence, Spanish determiner, Spanish noun. Error bars depict Standard Deviations.

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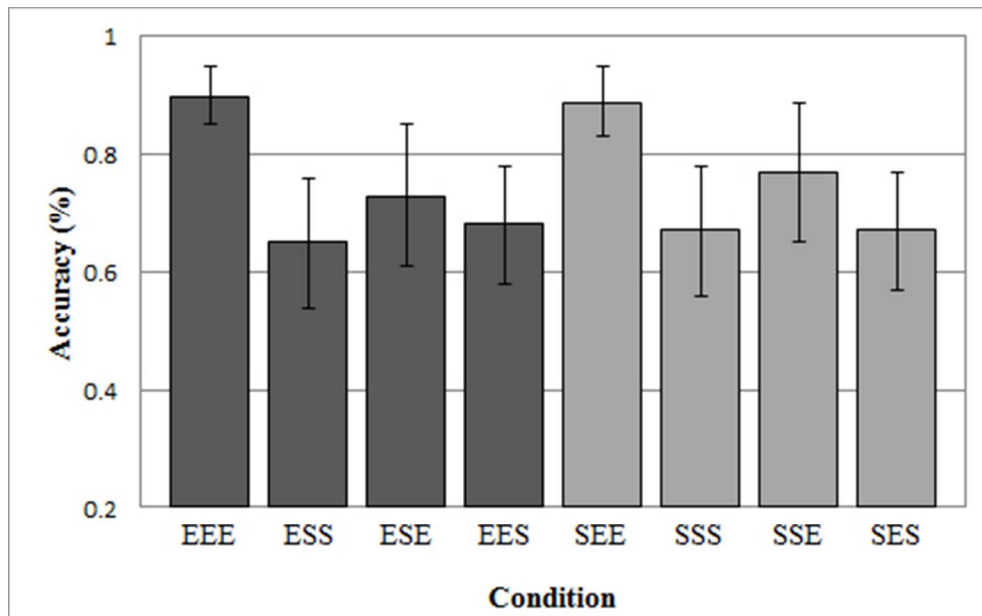


Figure 4. Mean Sentence Context Determiner-Noun Picture Naming Accuracy (Task 3). Dark bars indicate English sentence, light bars indicate Spanish sentence. EEE = English sentence, English determiner, English noun; ESS = English sentence, Spanish determiner, Spanish noun; ESE = English sentence, Spanish determiner, English noun; EES = English sentence, English determiner, Spanish noun; SEE = Spanish sentence, English determiner, English noun; SSS = Spanish sentence, Spanish determiner, Spanish noun. Error bars depict Standard Deviations.

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