



UNIVERSITY OF LEEDS

This is a repository copy of *Predicting language proficiency in bilingual children*.

White Rose Research Online URL for this paper:

<http://eprints.whiterose.ac.uk/151766/>

Version: Accepted Version

Article:

De Cat, C orcid.org/0000-0003-0044-0527 (2020) Predicting language proficiency in bilingual children. *Studies in Second Language Acquisition*, 42 (2). pp. 279-325. ISSN 0272-2631

<https://doi.org/10.1017/S0272263119000597>

© Cambridge University Press 2019. This article has been published in a revised form in *Studies in Second Language Acquisition* [<http://doi.org/10.1017/S0272263119000597>]. This version is free to view and download for private research and study only. Not for re-distribution, re-sale or use in derivative works.

Reuse

Items deposited in White Rose Research Online are protected by copyright, with all rights reserved unless indicated otherwise. They may be downloaded and/or printed for private study, or other acts as permitted by national copyright laws. The publisher or other rights holders may allow further reproduction and re-use of the full text version. This is indicated by the licence information on the White Rose Research Online record for the item.

Takedown

If you consider content in White Rose Research Online to be in breach of UK law, please notify us by emailing eprints@whiterose.ac.uk including the URL of the record and the reason for the withdrawal request.



eprints@whiterose.ac.uk
<https://eprints.whiterose.ac.uk/>

Predicting language proficiency in bilingual children

Cécile De Cat

Linguistics & Phonetics, University of Leeds

Abstract

Using advanced quantitative methods, this paper demonstrates that cumulative exposure to the school language is the best language experience predictor of proficiency in that language (as indexed by sentence repetition, lexical semantic and discourse semantic tasks) in a highly diverse group of 5- to 7-year-old bilingual children in monolingual education. An objective method is proposed to identify the amount of school language experience beyond which bilingual children are likely to perform within the monolingual range, and show that relative passivity in the home language does not translate into better school language proficiency. Socio-economic status is shown to interact in complex ways with language exposure, such that it is only above a certain level of exposure to the school language that the benefits of a more privileged background have a tangible impact on school language proficiency. To tease apart the effect of environmental predictors from the effect of cognitive factors, memory and cognitive flexibility measures are included as covariates in all analyses.

Predicting language proficiency in bilingual children

Introduction

Research on language proficiency in bilingual children has revealed a significant impact of language exposure, socio-economic status, and memory, although the interplay of all these factors has seldom been explored. Many studies tend to focus on predicting vocabulary, and until recently most research was based on relatively homogeneous home-language groups (although see Floccia et al., 2018 and Paradis, Rusk, Duncan, & Govindarajan, 2017).

While the impact of language exposure on language proficiency is uncontroversial, many questions remain about the relationship between the two. What aspects of language experience need to be taken into account? Is exposure alone a sufficient indicator? Does language *use* have a significant impact? Does past language experience matter (e.g., an initial monolingual period)? Is it possible to predict when bilingual children's proficiency in the school language can be expected to fall within the monolingual range, assuming the absence of developmental disorder?

Socio-economic status is also well established as an important environmental predictor, but how to interpret its effect (e.g., in relation to variations in quantity and/or quality of input) remains unclear.

This study offers an investigation of environmental predictors of language proficiency in bilingual children, informed by an in-depth review of the relevant literature and the use of advanced quantitative methods. As explained in the next section, the novelty of our approach lies in (i) the deliberate choice of a highly heterogeneous participant group, (ii) the breadth of proficiency aspects measured, (iii) the comparison of alternative estimates of language experience as predictors of language proficiency, (iv) the investigation of interactions between environmental predictors, and (v) the investigation of language exposure thresholds to inform comparisons between bilingual and monolingual groups. As such, our focus is both methodological and descriptive, so as to inform future research with a more theoretical focus.

The state of the art

Key aspects of bilingual language experience

Language development requires language input, i.e., direct evidence relevant to the acquisition of particular aspects of language. Language acquisition theories differ in what constitutes necessary input, and modelling the availability of usable input in the language environment is not straightforward (see Carroll, 2017 for discussion). It is however reasonable to assume that language exposure is a reliable proxy for language input. Indeed, variation in language exposure has been shown to affect many aspects of language development in bilingual children (see e.g., Grüter & Paradis, 2014; Hoff et al., 2012; Paradis, 2011; Paradis et al., 2017; Place & Hoff, 2011; Thordardottir, 2011; Thordardottir, Rothenberg, Rivard, & Naves, 2006; Unsworth, 2013b; Unsworth, Persson, Prins, & De Bot, 2014; Unsworth, 2017).

Bilingual children's language exposure is by its very nature extremely varied in terms of relative quantity and context (De Houwer, 2007). Some children are exposed to both languages at birth or thereafter, others are exposed to a second language later in childhood (De Houwer, 2011; Meisel, 2009). Relative exposure to the two languages may also vary over time, leading to different amounts of cumulative exposure to each over the course of their lives (Gutiérrez-Clellen & Kreiter, 2003; Thordardottir, 2011; Unsworth, 2013b). Given the heterogeneity of the bilingual language experience, it is now recognized that bilingualism should be conceptualized as a continuous variable rather than a categorical one (Kaushanskaya & Prior, 2015; Luk, 2015).

Studies that have investigated the relationship between language exposure and language proficiency in bilingual children have typically focused on a snapshot of the child's language experience at the time of assessment (De Houwer, 2009; G. Jia & Aaronson, 2003; Paradis, 2011). Recently, the notion of *cumulative* exposure has gained traction in the context of a more holistic approach to the child's language history (Gutiérrez-Clellen & Kreiter, 2003; Unsworth, 2013b). Indeed, proficiency at time x , however it is measured, is the result of the child's language experience over time and not just at time x . This is particularly relevant in the case of sequential bilingual

children who may experience rather abrupt changes in the relative exposure to their two languages when they start formal childcare placements or school.

The impact of bilingual children's language *use* on proficiency has hitherto received comparatively less attention (but see Bohman, Bedore, Peña, Mendez-Perez, & Gillam, 2010; Unsworth, 2015; Ribot, Hoff, & Burrige, 2017). Research with adult bilinguals has shown that the opportunity to use the language is an important factor in determining the degree of bilingualism (G. Jia, Aaronson, & Wu, 2002). There is reason to believe that, for children too, the extent to which they use a language will be a determinant of the proficiency attained in that language. Some studies with child bilinguals explicitly include a measure of language use, based on children's experience at the time of testing (Bedore et al., 2012; Bohman et al., 2010; Goldstein, Bunta, Lange, Rodriguez, & Burrows, 2010; Gutiérrez-Clellen & Kreiter, 2003). Similarly to language exposure, language use is likely to change over time as children's circumstances and their opportunities to use their two languages vary.

This study investigates how to best predict language proficiency based on language experience, by comparing different ways of operationalising language experience as a continuous measure. First, we ask if *a combined measure of cumulative language exposure and use* predicts (various aspects of) proficiency in the school/societal language. Then we compare that measure to alternative measures (cumulative exposure only, and current measures of exposure and use) to ascertain which is the most informative predictor in the population of interest.

Cognitive and social determinants of language proficiency

Memory skills, both Short Term Memory (STM) and Working Memory (WM) have been shown to be positively associated with language skills in monolingual children (Baddeley, Gathercole, & Papagno, 1998). STM is strongly associated with word learning skills (Avons, Wragg, Cupples Wragg, & Lovegrove, 1998; Majerus, Poncelet, Greffe, & Van der Linden, 2006), but the extent to which it is predictive of other language skills is less clear. Some studies have found a positive relationship

between STM and the *comprehension* of syntactically complex sentences in adults (Papagno, Cecchetto, Reati, & Bello, 2007; Lauro, Reis, Cohen, Cecchetto, & Papagno, 2010), others have failed to find such a direct link in children, in spite of a strong effect of STM on sentence *repetition* abilities (Willis & Gathercole, 2001). WM, on the other hand, is more clearly associated with metalinguistic abilities (McDonald, 2008) and sentence comprehension (Montgomery, 1995; Montgomery, Magimairaj, & O'Malley, 2008), particularly the comprehension of structurally complex sentences (see Kidd, 2013 for a recent review of the role of WM in children's sentence comprehension).

As for children who are exposed to more than one language, Thorn and Gathercole (1999) have shown a positive relationship between the ability to recall digits and non-words in two STM tasks and the receptive and expressive vocabulary skills of English-French bilingual between the ages of 4;0 and 8;0 in their two languages. STM as measured by Non Word Repetition has been shown to be more weakly associated with amount of exposure than vocabulary skills are (Thordardottir & Brandeker, 2013). To date very few studies have simultaneously investigated STM and WM in bilingual school-age children, and their relationship with lexical and grammatical skills. Engel de Abreu, Gathercole, and Martin (2011) explored this relationship in a group of Luxembourgish-German bilinguals between 5 and 6 years of age. STM was found to be predictive of vocabulary knowledge independently of WM, suggesting a highly specific association. WM and fluid intelligence (jointly interpreted as the capacity for controlled processing) were found to predict higher-order language processing (e.g., the understanding of syntactic contrasts). Compatible results were obtained by Verhagen and Leseman (2016), based on the comparison between a group of monolingual Dutch five-year-olds and an age-matched group of Turkish-Dutch bilingual children exposed to both languages naturalistically. The results of a confirmatory factor analysis showed that verbal STM predicted receptive vocabulary skills in Dutch, and that both verbal STM and VWM were predictive of grammar skills after controlling for vocabulary and SES.

While STM and WM can be considered child-internal determinants of language

acquisition, a child-external factor that has been shown to be of great significance is the socio-economic status (SES) of the child, indirectly measured via parental information. SES has been variously conceptualized in the literature in terms of household income, parental occupation or parental (often maternal) education. Given the typically high correlation between these variables, studies have often used them interchangeably.

The relative contribution of SES to bilingual children's language skills is not easy to identify. As highlighted by Hoff (2013), both bilingualism and SES have an impact on language development, and not enough research has yet considered the relative contribution of amount of language exposure and SES to disentangle these two factors. However those studies that have included bilingual children from different SES backgrounds (see below) have typically found that SES can be an independent predictor of language proficiency.

To date four studies have simultaneously evaluated the effect of SES and bilingualism on children's different linguistic and cognitive domains. Calvo and Bialystok (2014) focused on receptive vocabulary, non-verbal intelligence, and executive function tasks in a linguistically heterogeneous group of 6-7-year-olds in Canada with 26 different home languages. Gathercole, Kennedy, and Môn Thomas (2016) assessed vocabulary, grammar and cognitive skills in Welsh-English bilinguals in Wales. Chiat and Polišenská (2016) included Turkish-English and Spanish-English bilinguals in England and included measures of non-word repetition and vocabulary. Most recently Meir and Armon-Lotem (2017) addressed the independent and combined effect of SES and bilingualism in Russian-Hebrew bilinguals in Israel on expressive vocabulary and verbal short term memory.

The results for vocabulary measures are consistent across the four studies showing significant effects of SES and bilingualism, with bilingual children and children from lower SES performing more poorly. A lack of interaction between the two factors indicates that SES affects monolingual and bilingual children similarly.

Performance on a test of non-verbal intelligence in the Calvo and Bialystok (2014) study was not affected by either SES or bilingualism, while executive functioning was

adversely affected by low SES, but positively affected by bilingualism. In the study by Gathercole et al. (2016) SES, but not bilingualism, was a significant predictor of performance on cognitive measures.

As for memory, Chiat and Polišenská (2016) report that neither SES nor bilingualism affected children's performance on their non-word repetition tasks. By contrast, Meir and Armon-Lotem (2017) found that performance on a forward digit task was only affected by SES but not by bilingualism, even after controlling for vocabulary. For a NW repetition task neither SES nor bilingualism had an effect, with the exception of a negative effect of bilingualism on a subset of non-word-like stimuli which disappeared after controlling for vocabulary. In a sentence repetition task there was a significant effect both of SES and bilingualism, but the latter was non-significant after controlling for vocabulary.

The risk conferred by low SES is not induced by poverty itself, but by poverty of the child's environment, which translates into lower quantity and quality of the language addressed to the child. Factors such as high lexical diversity, syntactic complexity, and the frequency of decontextualized language use are all features of good quality input that facilitate the acquisition process, and they tend to be less well represented in parents from lower SES backgrounds (Hoff, 2006; Huttenlocher, Waterfall, Vasilyeva, & Vevea, 2010; Rowe, 2012).

This study will probe the impact of SES on language proficiency in two ways. Assuming SES is a proxy for the richness of the language environment, we will investigate how it interacts with the amount of language exposure experienced by the child. Also, given that both SES and cognitive abilities are predictive of performance in language proficiency tasks, and that SES predicts many aspects of cognitive development, we will investigate extent to which SES predicts language proficiency *over and above* the effect of cognitive predictors.

Bilingualism profiles

It is not uncommon to be faced with the question of “how bilingual” a child is. Parents of bilingual children who understand the home language but answer back in the societal language ask themselves if their child is “truly bilingual”. Teachers wonder which of the bilingual children in their care may need additional support with the school language. Researchers seeking to assess the impact of bilingualism on other aspects of a child’s development need to consider whether the children in their study are “bilingual enough” for an effect to be detectable — or may have as a primary question what the relevant threshold is.

Two questions are of particular interest. The first one relates to functional thresholds within the bilingualism continuum. The second one relates to language dominance more generally.

Functional thresholds. From what amount of language experience in a “second” language should a child be considered bilingual? In their study of vocabulary development in bilingual children, Pearson, Fernández, Lewedeg, and Oller (1997) report that children with less than 20% exposure to one of their languages tended “not to produce utterances in that language willingly or spontaneously” (p.56). While they were very careful to note that children below this exposure threshold might “become bilingual” nonetheless (and that some children with more exposure might not), their recommendation that the children in their sample “whose exposure was less balanced than 75:25 not be considered for future bilingual studies” (p.56) has been used as an exclusion criterion in many studies since (even if the use of that language was not itself under investigation, as in e.g., Laloï, de Jong, & Baker, 2017). The assumption is that, below a certain critical threshold of experience in more than one language, a child should be considered *functionally monolingual* (Bedore et al., 2012). What this means in practice is not clear. A functionally monolingual child may have a good level of comprehension in her “weak” language. Furthermore, while cross-linguistic interference during language comprehension (e.g., semantic priming) has been demonstrated in bilingual children as young as 30 months of age with at least 75:25 exposure balance

(Jardak & Byers-Heinlein, 2018), we do not yet know from what amount of language exposure the effect of cross-linguistic interference is detectable.

A related question is that of how much experience a child needs to have of a language to be likely to perform within the monolingual range of their age group. Many rightly object to the evaluation of bilingual children's performance in terms of deficiency in relation to monolingual norms, as language processing has been shown to be fundamentally *different* in monolinguals and bilinguals (Cook, 1994; Grosjean, 1989; Romaine, 1989). However, assessing whether a bilingual child's performance falls within the range of typically developing monolinguals (which itself features a considerable amount of variation) remains useful for education purposes, as bilingual children's academic achievement is assessed only in the school language. For instance, poor academic performance might be due to a lag in language proficiency rather than a lack in academic abilities. In turn, a lag in language proficiency will in most cases be due to lower levels of exposure to the school language. As the academic demands are the same for bilingual and monolingual children, it is important to be able to take into account the likelihood of a disadvantage induced by reduced exposure to the school language.

Cattani et al. (2014) provide evidence suggesting that typically-developing 2;6 year olds acquiring English and an additional language from birth will perform equivalently to their monolingual peers in terms of receptive and productive vocabulary if they are exposed to English 60% of the time.

What the threshold is for sequential bilinguals (who start out with a monolingual period) remains unknown. Also unknown is the threshold of language experience required for a child to perform within the monolingual range in other aspects of language proficiency. Our study will investigate these questions in 5- to 7-year olds, with respect to a range of proficiency measures.

Language dominance. Variation in bilinguals' language experience can result in an imbalance between their two languages. Children with superior proficiency in one language have been argued to be dominant in that language (following e.g., Genesee, Nicoladis, & Paradis, 1995), in contrast with "balanced" bilinguals, who are expected to

enjoy a similar level of proficiency in both languages. The notion of language dominance has prompted much controversy in the literature. Changes in language dominance over the lifespan (De Houwer, 2011) make it difficult to pin it down, and the distributed nature of bilingualism (where each language tends to be associated with different contexts — e.g., school vs. home) makes it difficult to assess language dominance across the board (Grosjean, 2016). There is also little agreement regarding how language dominance should be evaluated in the first place (see Silva-Corvalán & Treffers-Daller, 2015 for in-depth discussion).

However, there is a general consensus that language dominance is a useful construct, and that it can be estimated based on proficiency (Montrul, 2015) or based on language experience (Unsworth, 2015). A number of studies have also shown that language dominance predicts the directionality of cross-language influence in bilinguals: the dominant language has repeatedly been found to influence the non-dominant one (see e.g., Bernardini & Schlyter, 2004; Kupisch, 2007; Lanza, 2004).

While generally derived from continuous measures of language experience or proficiency, language dominance tends to be conceptualized in terms of discrete categories based on critical thresholds (e.g., Bedore et al., 2012). Yet, as argued by Treffers-Daller (2015, 253), language dominance should be regarded as a continuum, given that bilingualism is itself not categorical.

Our study will compare cumulative language exposure and language dominance as measures to estimate whether a bilingual child in monolingual education can be expected to perform within the monolingual range in terms of proficiency in the school language.

Summary. This review of the literature has brought to light a number of gaps which our study intends to address. These include (i) a systematic comparison of alternative measures of language experience (as predictor of a range of aspects of language proficiency) in bilinguals, (ii) a better understanding of the impact of socio-economic status, in relation with the amount of language exposure experienced by the child, and in relation with cognitive determinants of language proficiency (iii) the

objective identification of the amount of language exposure required for a bilingual child educated in monolingual settings to be expected to perform within the monolingual range in terms of language proficiency.

Aim and questions

Our overarching aim is to exploit advanced statistical methods to address the following three broad research questions, related to environmental predictors of language proficiency in bilingual children :

1. Do a gradient measure of bilingual language experience (combining exposure and use) and a gradient measure of socio-economic status (as a proxy for the richness of the child's language environment) significantly predict the following aspects of school language proficiency: comprehension and production of complex sentences, lexical semantics, and discourse-semantics?
2. How does a combined measure of cumulative language experience (combining exposure and use over time) compare with simpler measures (i.e., cumulative exposure or cumulative use, current exposure or current use), as predictors of school language proficiency?
3. What is the critical amount of school language experience required for bilingual children to perform within the monolingual range? In other words: at what point can one expect the gap between bilinguals and monolinguals to be "closed" in the school language?

We focus on children educated monolingually in the societal language (English), between the ages of 5 and 7 (i.e., in the first two years of formal education in England), with a broad range of exposure to a different language at home and diverse socio-economic backgrounds. Gender, age and cognitive factors (i.e., memory, cognitive flexibility) will be included in the analyses as covariates, as they can be expected to have an impact but are not central to our research questions.

Methodology

Participants

We aimed to study a representative sample of the bilingual population in a geographical area that is characterized by a high degree of linguistic and cultural heterogeneity. We recruited 174 children (including 87 monolinguals) between the ages of 5 and 7 from schools the North of England. Ethical approval was obtained from the University of Leeds (Ref. PVAR 12-007), and parental consent was obtained prior to data collection.

The school language was exclusively English for all the children. The bilingual children were also exposed to another language (henceforth the *home language*) in varying degrees (see below). There was a total of 28 home languages in our sample:¹ Arabic (9%), Bengali, Cantonese, Catalan, Dutch, Farsi, French (8%), Greek, Hindi, Italian, Kurdish, Mandarin, Marathi, Mirpuri, Nepalese, Pashto, Polish, Portuguese, Punjabi (21%), Shona, Somali, Spanish (6%), Swedish, Tamil, Telugu, Thai, Tigrinya, and Urdu (17%). Bilingual and monolingual children were recruited from the same schools for maximum comparability. None of the children were excluded from the study.

Table 1 summarizes the distribution of the two groups in gender and age. For ease of presentation, children with any amount of exposure to a language other than English are referred to as “bilinguals”; children who had no exposure to a language other than English are referred to as “monolinguals”. All children were reported by the school to be developing typically and did not have any known hearing deficit.

——- TABLE 1 HERE ———

Socio-economic profiles. The schools targeted were in areas of varying degrees of affluence, so as to recruit children from as broad as possible a socio-economic spectrum. The socio-economic status of the children’s families was estimated on the basis of information gathered via a parental questionnaire (see Appendix). Two measures were obtained: one for parental level of education (1), one for parental level of current occupation. The highest level was chosen in each case (on the assumption that the status of the household was determined by the best educated parent and the highest

occupation among the parents).

- (1) a. None
- b. Primary school
- c. Secondary school
- d. Further education
- e. University

The occupational data was scored using the reduced method of the National Statistics Socio-economic Classification (simplified NS-SEC, which is based on the Goldthorpe Scheme of sociological classification (Goldthorpe, 1980).² The score obtained was reversed for ease of interpretability (i.e., a positive correlation with educational level).

The two measures are significantly associated ($\chi^2(4, N = 174) = 83.57$, $p < 0.0001$), as shown in Figure 1. There is a weak but significant negative correlation between the cumulative amount of home language exposure and socio-economic status (based on the occupational classification): $r = -0.24$ $p = 0.0014$.

————— FIG. 1 HERE —————

Language experience. Estimates of the amount of exposure and use for each child in English and in their home language were calculated on the basis of information gathered via parental questionnaires (see Online Supplement).

The information sheet, consent form and questionnaires were translated into the three most common home languages with a high risk of low English proficiency in the parents (Bengali, Punjabi, Urdu). Because some of the parents speaking these languages turned out to be illiterate, help was offered either from the research assistants, the teachers or other employees at the schools, or in some cases from other parents who were able to translate.

The language questionnaire asked parents about their own proficiency in English and any other languages,³ and their child's exposure and use of English and their other language in a range of contexts inside and outside the home. It was based on a simplified version of the BiLEC (Unsworth, 2013b), to make them understandable by

parents with limited levels of literacy, without one-to-one support for filling the questionnaire (as this was not feasible for our sample size, given our means and time-frame). Parents were asked to list their child's interlocutors at various time periods during weekdays and during weekends in school term, as well as during holidays. The time periods captured all the child's waking hours. School hours corresponded to a single block and the rest of school days was broken down by hours. Weekends and holidays were broken down in time slots of two hours. Type of interlocutor was constrained as a choice between mother, father, siblings, school, or other. For each interlocutor (except 'school'), we asked which language was used to address the child, and which language the child used with that interlocutor.⁴ The language of interaction had to be specified as either *English, home language, other home language* (i.e., other than the main one identified at the outset of the questionnaire), *2 home languages*, or *both English and the home language*. For each interlocutor (including the child), the proportion of interaction in English or the home language had to be specified (as *always, usually, half of the time, rarely* or *never*, which we converted into a 5-point scale ranging from 100% to 0%). For mother, father, and any other significant carer in the home, their proficiency in English was rated by the parent filling the questionnaire (as speaking the language *very well, quite well, not well* or *not at all*). Parents also reported the age of onset of exposure to English, the context of that first exposure (*home, play group, nursery, school* or *other*) and whether their child was born in the UK. The total number of weeks spent in the home-language country was also reported. Finally, parents were asked to indicate the types of activities their children engaged in, in each language (*reading with an adult, using a computer, watching television, sports/club, playing with friends*).

Current measures of language exposure and use were calculated as follows. The total number of hours of interaction with the child per year was calculated for each of the child's interlocutors. The proportion of total exposure received by the child from each interlocutor in the home language was calculated as the total number of hours of interaction with that interlocutor, multiplied by the proportion of the time the home

language was used with that interlocutor. By adding these proportions from all the child's interlocutors and dividing the sum by the total number of hours of interaction, we obtained a measure of current exposure to the home language for that child, expressed as a proportion of their total interaction time (assumed to equate to waking hours). Current exposure to English was calculated in the same way. The sum of current exposure to the home language and in English came to 100%. We repeated the same procedure for language use.

Cumulative measures of exposure and use of the home language were calculated as the sum of (i) the number of months of monolingual exposure (i.e., prior to onset of exposure to English — this amounted to 0 for simultaneous bilingual children and for monolingual children) and (ii) the number of months of bilingual exposure multiplied by the proportion of current exposure to (or use of) the home language. The cumulative measures thus correspond to the total number of months equivalent to full-time exposure to the home language.

The cumulative measures do not make a distinction between stages of linguistic development. One might object that, during the earliest months of life, the amount of exposure to each language might be of little consequence, given that the child is not yet able to produce use of a particular language. However, it has recently emerged that the first year of life constitutes a critical period for the development of syntax (Friedmann & Rusou, 2015). We therefore take the default assumption to be that exposure to a language has an impact from the earliest stages of exposure.

A limitation of our estimation methods is that the cumulative measures cannot take into account variability over time, except that induced by age of onset of exposure to the societal/school language.

Our sample ranges across an evenly distributed continuum of bilingual language experience. At the lower end, some children had had a very limited experience in a language other than English. At the higher end, some children were late bilinguals, having only experienced the home language until their first significant exposure to English at primary school.

As shown in Figure 2, in the bilinguals, current exposure to the home language ranged from 9% to 89% of estimated waking hours (left panel) while cumulative exposure to the home language ranged from 6 to 69 full months equivalent (right panel). At the time of the study, 31 (36%) children used their home language to the same extent as they were exposed to it (as shown by the top diagonal line in the plot). The rest of the children were relatively more “passive” in their home language, with 9 children totally “passive” over their lifetime (i.e., always answering in English when addressed in the school language).

It is worth noting that simultaneous bilinguals, who were exposed to the two languages from birth, are not necessarily less “passive” in the home language: out of the 41 simultaneous bilinguals in our sample, only 11 used their home language to the same extent that they were exposed to it at the time of the study. Furthermore, even if a child started out as monolingual in the home language, this is no guarantee that they will remain “active” in that language when they start school in English. For instance, out of the 18 children who had had very little exposure to English before the age of 3, eight of them had become relatively “passive” in the home language at the time of the study.

————— FIG 2 HERE ———

Children who are bilingual from birth did not necessarily come from “one-parent one-language” households (see Table 2): only 51% (21/41) of simultaneous bilinguals did so in this study. English exposure started outside the home (in daycare) for 17% (7/41) of the simultaneous bilinguals. Given this variability, neither the language policy of the household nor the age of onset of bilingual exposure are sufficient to give the full picture of children’s patterns of language exposure over time.

Proficiency measures

Several measures of English language proficiency were collected from a battery of tests aiming to tap into a broad range of aspects of language competence. The tests included (i) a sentence repetition task (the short version of the LITMUS Sentence Repetition test — Marinis, Chiat, Armon-Lotem, Gibbons, & Gipps, 2010; Marinis &

Armon-Lotem, 2015), (ii) the sentence structure sub-test of the Clinical Evaluation of Language Fundamentals (CELF-4-UK — Semel, Wiig, & Secord, 2006), and (iii) four sub-tests of the Diagnostic Evaluation of Language Variation: the verb and preposition contrasts, real verb mapping, novel verb mapping and articles sub-tests (DELV — Seymour, Roeper, & de Villiers, 2005). The aspects of proficiency probed by each test are outlined below.

The sentence repetition task (SRep) was designed specifically for bilingual populations.⁵ It taps into language processing at all levels of representation (phonological, morpho-syntactic and semantic) in both comprehension and production, as successful comprehension is necessary for correct repetition (Marinis & Armon-Lotem, 2015). We used the short version, which comprizes 30 sentences with three levels of difficulty described in Table 3.

———— TABLE 3 HERE ————

The Sentence Structure component of the CELF aims to “measure the acquisition of grammatical (structural) rules at the sentence level” (CELF manual p.88). In each trial of this comprehension test, the child is presented with four pictures and asked to point at the one that matches the prompt sentence. The test is designed for 5- to 8-year-old children.

The DELV is a dialect-neutral assessment for 4- to 9-year-olds, aiming to limit the effects of language exposure differences in bi-cultural populations. We administered the following tasks to evaluate children’s semantic competence, and in particular lexical semantics, argument structure, and discourse semantics:

- (2) a. Verb contrast and preposition contrast
- b. Real verb and novel verb mapping
- c. Articles

The contrast tasks tap into the organization of the child’s lexicon into contrastive words and levels of meaning. These tasks focus on action and location words, which relate to common aspects of children’s experience across cultural/social groups. They require the

child to provide a word that contrasts appropriately with the one negated in a prompt sentence, based on the information provided by the accompanying picture. For instance, in the preposition contrast task, the child might be presented with a picture showing two children pulling angrily at the same train and with the prompt in (3). Adequate answers are illustrated in (a), and inadequate ones in (b).

- (3) They're not sharing the train, they're...
- a. e.g., fighting over it, arguing over it (ADEQUATE RESPONSES)
- b. e.g., fighting, breaking it (INADEQUATE RESPONSES)

The preposition contrast task comprises 6 items, and the verb contrast task comprises 10 items.

In the mapping tasks (2-b), the child is presented with a series of pictures representing an event, and a second series of side pictures representing the participants in the event (see Figure 3). The examiner describes the event (as in (4)) while pointing at the pictures in a continuous, flowing motion, and then asks the child a series of questions (4-a)-(4-d).⁶ The child answers by choosing one of the side pictures.

- (4) The boy is pouring the juice.
- a. Which one was the pourer? (boy)
- b. Which one got poured? (pitcher of juice)
- c. Which one did he pour the juice into? (glass)
- d. Which one as pourable? (pitcher of juice)

The novel verb mapping task follows the same protocol, but with a nonce verb. The aim of this task is to assess children's abilities to map new meanings from the linguistic context onto verb frames.

————— FIG. 3 HERE —————

- (5) The man is lelling the clown.

- a. Which one was the leller? (man)
- b. Which ones was lellable? (clown)
- c. Which one was the clown holding on to? (DISTRACTOR)
- d. Which one got lellled? (clown)
- e. Which one was lelling? (man)

The real verb and the novel verb mapping tasks comprize three verbs/scenarios each, about which the child is asked 10 and 15 questions respectively.

Finally, the Articles test of the DELV (2-c) aims to assess children's understanding (and implementation) of the discourse-semantic differences in the English article system. Children were asked a series of 30 questions (each preceded by a minimal verbal context) requiring a noun phrase as their answer. Based on their assessment of what the interlocutor knows, the child has to choose between a definite (*the*) or an indefinite (*a*) article for that noun phrase (Schafer & de Villiers, 2000). This depends on how the object is defined and what is known or assumed about the object by the listener. The test was preceded by practice trials until it was clear that the child understood the task.

- (6) George bought a shirt and a bag. He's wearing one of them. Guess which.
- a. the + appropriate noun (e.g., the bag or the shirt) (CORRECT)
 - b. a + appropriate noun (INCORRECT)
 - c. bare noun; inappropriate noun; no response (INCORRECT)

Cognitive measures

Language proficiency tests require children to process and remember language prompts but also, when based on pictorial stimuli, to compute inferences (e.g., to choose the correct picture from a set). We therefore collected several cognitive measures to use as covariates in the analyses.

Measures of short-term and working memory were obtained from the Digit Span tasks (Wechsler Intelligence Scale for Children III — Wechsler, 1991). The use of numerical memoranda has been shown to be relatively independent of test language and

cultural status (Engel de Abreu, Baldassi, Puglisi, & Befi-Lopes, 2013). In those tasks, the examiner verbally presents digits that the child has to repeat in the same order (in the Forward Digit Recall task) or in reversed order (in the Backward Digit Recall task). The number of digits increases by one until the child consecutively fails two trials of the same digit span length. There is a maximum of four trials per digit span length.

The Forward Digit Span measure was used as a proxy for children's short term memory capacity, a key component in Baddeley's (2000) Multicomponent Working Memory Model, which has been argued to represent a constraint in language processing (Boyle, Lindell, & Kidd, 2013). The Backward Digit Span measure was used as a proxy for children's working memory capacity, which has been shown to correlate with spoken sentence comprehension (Montgomery et al., 2008; Magimairaj & Montgomery, 2012). The raw results on the memory tests are summarized in Tables 4 and 5.⁷

————- TABLE 4 HERE ————

————- TABLE 5 HERE ————

To assess children's cognitive flexibility, we administered the Dimensional Change Card Sort task (henceforth DCCS). The protocol was as described in Zelazo (2006). The child was presented in each trial with a picture representing either a rabbit or a boat, that was either blue or red, and asked to place the card in one of two boxes according to a sorting rule (by shape or by color). The boxes were identified by either a blue rabbit or a red boat. The first block trials (N= 6) required using the shape criterion, and the second block (N= 6) required using the color criterion. As the children in our sample were older than 5, we also administered a more advanced block of trials (N= 12) in which a star appeared on some of the cards. Cards without a star had to be sorted according to the shape dimension, and cards with a star had to be sorted according to the color dimension. The repetition of instructions on every trial (in all blocks) ruled out the possibility that difficulty could be attributed to hypothesis testing or memory of the relevant rules. The test session was preceded by a demonstration and two practice trials. The test trials were presented in two counterbalanced orders.

We adopted the method recommended by Zelazo (2006) to score the DCCS data.

This is based on a pass-fail criterion on each trial block. Passing the first two blocks requires sorting at least 5 out of 6 cards correctly (on each block). Passing the third block requires sorting at least 9 out of 12 cards correctly. A child is assigned a score of 0 if they fail the pre-switch block, a score of 1 if they pass the pre-switch block only, a score of 2 if they also pass the post-switch block, and a score of 3 if they pass all three blocks. Little variability is observed, as shown in Table 6, as many children performed at ceiling. Only one (bilingual) child did not pass the first block.

————— TABLE 6 HERE —————

Descriptive results for the proficiency tests

Results for each proficiency test are first reported descriptively, showing the range of performance of bilinguals compared with that of monolinguals. Unsurprisingly, as a group, the bilinguals' performance is below that of monolinguals in each test.

Sentence repetition

The SRep test can be analyzed in different ways, depending on the criterion against which the children's utterances are evaluated (Marinis & Armon-Lotem, 2015):

- (7) a. exact repetition of the target sentence (Full Accuracy)
- b. grammatical production of the targeted structure (Target Accuracy)
- c. grammaticality of the sentence produced (whether or not it fully matches the target) (Grammaticality)
- d. types of mismatches (i.e., omissions, substitutions or additions) compared with the target sentence) in the following categories:
 - (i) Lexical
 - (ii) Inflectional
 - (iii) Functional

The distribution of raw scores according to the first three types of analysis is plotted in Figure 4. Utterances containing unintelligible material were excluded (total: 208, i.e.,

4% of the data).

————— FIG. 4 HERE —————

Sentence comprehension

Although not designed for bilingual populations, the CELF is frequently used to evaluate the proficiency of bilingual children (e.g., Paradis, Crago, Genesee, & Rice, 2003; Barac & Bialystok, 2012).

The distribution of mean scores in the CELF Sentence Structure task is shown in Figure 5.

————— FIG. 5 HERE —————

Lexical-semantic

We analyze the lexical-semantic tests of the DELV separately from the discourse-semantic test, as they tap into different domains of language competence: The lexical-semantic tests tap into the organization of the lexicon, and the discourse-semantic tests taps into the child's ability to express contrasts in information status through the appropriate use of determiners.

The distribution of raw scores on the lexical-semantic tests is plotted in Figure 6.

————— FIG. 6 HERE ———

Discourse-semantic

The distribution of the mean scores for the Articles test of the DELV are plotted in Figure 7.

————— FIG 7 HERE —————

Methods of analysis

To model the relationship between English proficiency (as indexed by the response variables from each test) and the potential predictors of proficiency we aimed to investigate, we performed linear regression analyses. For each predictor, the model effectively fits a single line through the scatterplot of individual scores, showing how the

mean response changes with the predictor variable. The model can also account for the interaction of two (or more) predictor variables. An advantage of regression analyses is that they can model variance due to continuous as well as categorical predictors.

Linear models assume that the observations (of the response variable) are independent. However, in each proficiency test, multiple observations were collected from each individual, resulting in a nested data structure (rather than truly independent observations). To account for this, and to control for sampling effects (due for instance to lexical properties of the test items), we included Participant and Test Item as random effects in each model, alongside the fixed effects (the predictor variables). The resulting analysis is a *mixed effect* model, as it combines random and fixed effects in a single analysis (Baayen, 2008).

We fitted linear mixed models to the data from each proficiency test, using the ‘lme4’ package (version 1.1.21) in R (version 3.6.0). The models were built by adding predictors incrementally, starting from a null hypothesis model including only random effects for Participant and Item. A predictor was retained only if it improved the fit of the model, yielding a significant reduction in AIC⁸ and a significant R-squared value for the model.⁹ Random slopes lead to model non-convergence and were therefore not included in the final models. In each case the final model was checked against the null hypothesis model, and its residuals (i.e., the observations not accounted for by the model) were checked for normality of distribution, as a further test of the quality of the model.

In all analyses, we tested whether the following variables were significant predictors: age, home language experience, SES, short-term memory, working memory and gender (treating the latter three as covariates). All continuous variables were standardized so that the impact of the predictors can be compared within each model.¹⁰

For each analysis, we report the optimal model in a table. The statistics for non-significant factors will be given in the text where relevant. In the first instance, we present models identifying the predictors of proficiency in bilingual children only. Subsequently, we include monolingual children in the models and identify the threshold

of home language experience below which the performance of “bilinguals” does not differ significantly from that of monolinguals in the language of schooling.

Question 1: Predicting language proficiency from bilingual children’s language experience and SES

Before reporting on the analyses carried out to address this question, we explain how language experience was apprehended as a variable.

We saw that children’s experience of their home language (HL) changed over time, and that many tended to become more “passive” in their usage (even if they spoke only the HL before the age of 3 or 4). Cumulative measures of language experience are therefore likely to be more accurate than current measures, as they can reflect these changes over time. Consequently, we based the first set of analyses on cumulative measures (and then subsequently compared them with analyses based on current measures, to see which fared better).

As all the children in our sample were exposed to two languages only, the amount of experience obtained in one language was inversely correlated with that obtained in the other language (e.g., cumulative exposure to HL vs. English: $r = -0.955$, $p < .0001$). In terms of statistical modeling, it should therefore not make a difference whether exposure to one or the other language is used as a predictor. For practical reasons, we chose to base our language experience measure on the HL measure, so as to make it more easily interpretable: Monolingual children would naturally get a bilingual score of 0 whatever their age, as they had no exposure to a HL other than English; for bilingual children, the higher the score, the greater the child’s experience in their HL (and the lesser their experience in English). Another advantage of basing our combined measure on the HL is that, given the substantial variation in HL experience in the group under investigation, the resulting measure is not correlated with age. This makes it possible to tease apart the respective effect of language exposure and age within a single model.

To take “passivity” into account (on the assumption that a bilingual who is relatively passive in a language is likely to be more proficient in their other language),

language use has to be taken into account in relation with language exposure.

Regression analysis can deal with the effect of continuous predictor variables, but is not able to handle severe collinearity between predictors — as is the case with our language exposure and use measures.¹¹ We therefore combined the cumulative measures of home language exposure and use into a single measure, as explained in the next section.

To address the first part of our first research question (i.e., whether a gradient measure of language experience (combining exposure and use) significantly predicts school language proficiency), we created a new language experience measure combining the cumulative estimates of language exposure and use. Because English is the language of schooling, we assumed it was actively used by all children. Greater variability was expected in the exposure to and use of the home language (HL), which, for many children, was (or would become) the weaker language. We therefore used HL estimates to derive our combined language experience measure.

As will be shown below, this combined measure aligns very closely to the amount of cumulative HL exposure, with a “correction” if the HL is productively used by the child to a lesser extent than they are exposed to it (i.e., if cumulative use is lower than cumulative exposure). It can therefore be conceived as a measure of cumulative HL exposure, adjusted for relative “passivity” in that language. We will refer to this combined measure as the Bilingualism Profile Index (BPI).

The BPI was derived by a standard measure of dimensionality reduction: a Principal Component Analysis of the two scores.¹² The choice of this method to combine the two scores is justified as follows. The apparently simpler alternative, i.e., linear combination of exposure and use measures, would require arbitrary manipulation, as a particular mathematical operation (e.g., multiplication, division) is required. The choice of operation should be justifiable on conceptual grounds, such that the resulting measure is clearly interpretable. Multiplying exposure and use or dividing one by the other would not result in scores that could be meaningfully interpreted. By contrast, dimension reduction (here by Principal Component Analysis — henceforth PCA) does not encounter these conceptual difficulties: the combination of the two measures is

entirely determined by their distributional properties and exempt of (arbitrary) intervention from the researcher. The resulting score simply combines the information contained in the original two scores. The BPI can thus confidently be interpreted as a score for home language exposure “corrected” downward in the case of comparatively lower use of the home language (see below for illustrations).

The PCA of cumulative HL exposure and use yields two principal components, the first of which accounted for most (98%) of the variability in our dataset (due to the strength of the correlation between the two cumulative measures). The first component thus contains almost all the information from the two original measures. The BPI scores correspond to the reversed loadings of that first component.¹³ The monolingual children obtain a BPI score of 0, as their score was 0 for both cumulative measures of HL experience. The bilinguals’ BPI scores ranged from 4 to 96.

Figure 8 shows the correlation of the BPI with current and cumulative measures of exposure and use of the home language.¹⁴ The dispersion from the linear relationship between the BPI and each of the cumulative measures shows that it is not reducible to any of these measures.¹⁵

————— FIG. 8 HERE —————

The BPI is very strongly associated with the age of onset of exposure to English ($\chi^2 = 6876.01$ $p < 0.0001$), but simultaneous bilinguals do not necessarily have less home language experience than sequential bilinguals (who experienced an initial monolingual period in that language). This is shown in the left pane of Figure 9.

The BPI is also associated with the type of home language environment, which we operationalize as follows:

(8) Home language policy categories:

HS (“home split”): one-parent one-language

HOL (“home only, low”): both parents speak the home language; siblings tend to speak English

HOH (“home only, high”): the whole household speaks the home language

While Bilingualism Onset and Home Environment are significantly correlated or associated with the BPI score, neither of them is able to capture the variability in HL experience as precisely as the BPI. For instance, Figure 9 shows that there was a substantial amount of variability in the amount of HL language experience obtained by children who were bilingual from birth.

————— FIG. 9 HERE —————

The BPI has two methodological advantages. As a *single* measure capturing cumulative exposure and use, the BPI makes it possible to use the combined impact of these two highly correlated measures as a predictor in regression analyses.

To answer to our first research question, we investigate below the impact of the BPI and SES as predictors of children’s proficiency in English (the school language) in the age group under consideration. The proficiency measures we consider in turn are: sentence repetition, sentence comprehension, lexical semantics, and discourse-semantics.

Sentence repetition

Using the methodology described above, we fitted a mixed-effect model to bilingual children’s accuracy data. The response variable was the correct repetition of the target structure on each of the SRep test items (see (7-b)). It indexes children’s ability to produce structures of increasing grammatical complexity (listed in Table 3) independently of their ability to remember exactly all the words in the sentence.¹⁶

Table 7 summarizes the effect of the significant predictors of accuracy of the production of the target structure by the bilingual children (expressed in our model as the probability of reproducing that structure correctly), as per the optimal model. The Estimates in the model summary report the likelihood of a correct response¹⁷ for each predictor, compared with the likelihood observed at the Intercept. As all the continuous predictors were transformed into z-scores, the intercept is based on the combination of the mean values: the average score for SES, memory, age and BPI characterizing the bilingual group. In terms of categorical variables, the intercept is calibrated for SRep Level 1 and Gender:Female.

As seen in the model summary, the BPI is the strongest child-related predictor ($\beta = -0.7026, p = 0.0008$). Children's ability to reproduce the target structure is negatively correlated with the amount of experience in their home language: a higher BPI score decreases the probability of a correct response significantly. Children's performance was also predicted by their socio-economic status ($\beta = 0.5842, p = 0.002$) and their short term memory ($\beta = 0.5556, p = 0.006$). The effect of age (over and above that of short term memory) also approached significance ($\beta = 0.4082, p = 0.069$). Structural complexity (SRep Level) had a significant impact on performance. The more complex the targeted structure, the poorer the performance: even Level 2 yielded significantly poorer results than Level 1 ($\beta = -0.9938, p = 0.041$).

There was no interaction between structural complexity level and children's amount of bilingual experience ($\beta = 0.13, p = 0.45$). Working Memory had no significant effect $\beta = -0.05, p = 0.77$). The random effects for Participant and Item had a variance of 2.55 and 1.12 respectively.

The same predictors were found to be significant in the other overall accuracy analyses of the bilingual children's performance in the SRep test (not reported here). The proportion of variance captured by STM compared with the other predictors is greater in the Full Accuracy analysis. In the Grammaticality analysis, the level of difficulty (SRep Level) is a lower-ranking predictor (which is to be expected as this analysis does not penalize avoidance strategies), with no significant difference between complexity levels 1 and 2. The relative ranking of predictors by decreasing order of importance in each model are listed in (9) (where SRep Level stands for the items' level of difficulty):

- (9) a. Full Accuracy (7-a):
STM > SRep Level > SES > BPI
- b. Grammaticality (7-c):
STM > SES > BPI > SRep Level

Error analysis was performed according to the domain affected (functional,

inflectional, lexical). Errors are illustrated below for the functional domain (10), the inflectional domain (11), and the lexical domain (12) respectively.

- (10) Target: He will feed the cow before he waters the plants.
 He will feed the cow when he waters the plants. (Functional error)
- (11) Target: The children were taken to the office.
 The children taked to the office. (Inflectional error)
- (12) Target: The mum bakes the meal that the children are eating.
 The mum makes food for the children. (Lexical errors)

Three accuracy scores were obtained for each sentence by dividing the number of errors of each type from the total number of relevant words (Marinis et al., 2010).

We tested the significance of the same predictors as in the above analyses. The full models are reported in the Online Supplement. The relative importance of predictors (and their statistical significance) differs across error types, as shown in (13) (non-significant predictors are omitted).

- (13) Functional: SRep Level > BPI > STM > SES
 Inflectional: SRep Level > STM
 Lexical: SRep Level = STM > SES > BPI

Neither the BPI nor socio-economic status turn out to be significant predictors of inflectional errors. The relative importance of the BPI is highest in the model predicting functional errors. The relative importance of short term memory is highest in the model predicting lexical errors.

Sentence comprehension

A linear regression analysis of the CELF Sentence Structure data reveals that bilingual children's scaled scores were *not* predicted by children's BPI ($\beta = -0.18$, $p=0.58$), nor their socio-economic status ($\beta = 0.51$, $p=0.09$), or their short-term

memory ($\beta = 0.14$, $p=0.67$). Instead, the strongest predictors were working memory ($\beta = 0.58$, $p=0.03$) and cognitive flexibility, ($\beta = 1.36$, $p=0.00001$).

In (De Cat, under review), I argue that the CELF Sentence Structure test does not probe children's ability to deal with linguistic aspects of structural complexity, as these are confounded with cognitive complexity in the task. The lack of significance of English exposure (as the counterpart of HL exposure) as a predictor is most likely due to that confound. It is unclear whether the Sentence Structure scores can be interpreted as a reliable index of language proficiency. We will therefore not consider them any further in this paper.

Lexical-semantics

Given the different distribution of scores across tests in the bilingual group (Welch's $F(3, 190.795) = 31.0695$, $p < 0.0001$), we did not use a composite score in the first instance. We carried out a MANOVA to identify the significant predictors of performance in the four DELV sub-tests (in one go) without inflating the Type 1 error rate.

Using Pillai's trace, a statistically significant MANOVA effect on the performance of bilingual children was observed for the BPI ($V = 0.2$, $F(4, 77) = 4.7$, $p = 0.002$), socio-economic status ($V = 0.18$, $F(4, 77) = 4.23$, $p = 0.004$), short term memory ($V = 0.18$, $F(4, 77) = 4.36$, $p = 0.003$), and cognitive flexibility ($V = 0.23$, $F(4, 77) = 5.77$, $p = 0.0004$). Age approached significance ($V = 0.11$, $F(4, 77) = 2.38$, $p = 0.0586$), and working memory was not significant ($V = 0.063$, $F(4, 76) = 1.28$, $p = 0.28$).

After the MANOVA had identified the significant predictors, linear regression models were fitted to predict the accuracy score on each of the DELV sub-tests. The BPI was a significant predictor of poorer performance in the Preposition Contrast task ($t(82) = -2.63$, $p=.01$) and the Real Verb Mapping task ($t(82) = -2.39$, $p=0.02$), and approached significance in the Verb Contrast task ($t(82) = -1.80$, $p=0.07$). It was not significant in the Novel Verb Mapping task ($t(82) = -0.81$, $p=0.42$). SES significantly predicted performance in the Verb Contrast task ($t(82) = 2.57$, $p=0.01$), and

approached significance in the Real Verb Mapping task ($t(82) = 1.96, p=0.05$). It was not significant in the Preposition Contrast task ($t(82) = 1.47, p=0.14$) nor in the Novel Verb Mapping task ($t(82) = 1.37, p=0.17$). Short term memory significantly predicted performance in the Real Verb Mapping task ($t(82) = 4.26, p < 0.0001$) and the Novel Verb Mapping task ($t(82) = 3.32, p=0.001$). It approached significance in the Verb Contrast task ($t(82) = 2.56, p=0.1$), and was not a significant predictor in the Preposition Contrast task ($t(82) = 1.47, p=0.14$). This is summarized in Table 8.

———— TABLE 8 HERE ————

If the four scores are combined into a composite score, the following factors are found to predict bilingual children's performance (in order of importance), as summarized in Table 9: short term memory, BPI and working memory. SES and Age are not significant.

Discourse-semantics

Analysis was performed by fitting a linear regression model to the children's overall score. Performance was negatively correlated with the BPI ($\beta = -0.0485, p=0.012$). This was more than compensated by SES, which was a stronger predictor of performance in the bilingual children ($\beta = 0.0549, p=0.0074$). The following factors proved non-significant predictors of the bilingual children's performance: age ($\beta = 0.0306, p=0.12$), short term memory ($\beta = 0.0125, p=0.53$) and working memory ($\beta = 0.029, p=0.14$). Table 10 reports the summary of the optimal model.

Summary

In answer to our first research question, we have demonstrated that the BPI, as a gradient measure of language experience combining language exposure and use, is a significant predictor of several aspects of school language proficiency in the population under study. The BPI is the strongest child-related predictor in the sentence repetition test (if scored based on the correct repetition of the target structure rather than full accuracy). It is also a strong predictor in the lexical semantic tests (the strongest predictor after cognitive flexibility) and in the discourse semantic test. Although the

BPI indexes cumulative HL exposure and use in our models, the interpretation of its effect should not be that HL experience has a detrimental effect. The BPI was used as an indirect estimate of the amount of experience in the school language (as it is the inverse picture of HL experience in bilinguals). Therefore our findings are in line with much previous research showing a positive association between amount of language input and use to performance on tests in the school language.

Furthermore, bilingual children's socio-economic status is a significant predictor of performance in the sentence repetition task and in the Discourse-Semantics tasks. It was found not to be a robust predictor in the Lexical Semantics tasks.

Children's cognitive abilities were included as a covariates, and found to have an impact in all tasks (memory in most tests, and cognitive flexibility in those tests relying on inferencing abilities). Performance in the sentence comprehension task (the Sentence Structure task from the CELF) was correlated with cognitive predictors only, which we interpret as a problematic confound (see De Cat, under review for substantiation).

Question 2: Comparing alternative measures of language experience

Having demonstrated that a gradient, composite measure of language experience significantly predicts several aspects of bilinguals' proficiency in the school language, we now turn to our second research question, which seeks to identify the optimal way of modelling language experience as a predictor of school language proficiency. We break it down into the following two questions, which we address in turn below:

1. How does a combined measure of language experience (combining exposure and use) compare with simpler measures (i.e., exposure *or* use), as predictors of school language proficiency? Are *cumulative* measures more accurate than current measures, in predicting school language proficiency?
2. What is the specific impact of language *use* as a predictor of school language proficiency?

Question 2.1: Cost-benefit analysis: which language experience measure to use as a predictor?

The BPI indexes the cumulative amount of exposure and use of the home language (HL), and it is strongly correlated with alternative measures of language exposure (see Figure 8). Our rationale in choosing the BPI as predictor over these alternative measures was based on a number of considerations: (i) Calibrating language experience on the HL makes it possible to include monolinguals in the continuum “naturally” (as they have had zero experience of a HL different to English); (ii) Encompassing language use (as well as exposure) yields a more precise measure, which in turn should increase the power of our analyses; (iii) Focusing on cumulative experience further increases the accuracy of the measure, as it reflects the quantitative effect of sequential bilingualism.¹⁸ It is however possible that the higher level of precision of the BPI does not afford a substantial benefit over a simpler measure. In this section, we compare the BPI with simpler, alternative measures of language experience: cumulative exposure, cumulative use, current exposure, and current use.

To determine the optimal language experience measure predicting school language proficiency, we exploited the method described in Burnham and Anderson (2003): an information-theoretic approach to inform model selection and multi-model inference. Choosing between alternative statistical models requires estimating which of them best approximates the “true” process underlying the phenomenon under study. If several models compete closely, the “true” process should be inferred by combining the information from these models. In this approach, both processes are informed by the *Akaike weight* associated with statistical models.

The Akaike weight of a model is based on the Akaike Information Criterion (AIC), which was defined above as an indicator of the trade-off between the accuracy and the complexity of a given model (i.e., a measure of the relative goodness of fit of the model to reality). When considering two nested alternative models (i.e., models that differ only in the presence or absence of one or more variables), the model with the lowest AIC should be chosen as the best model.

By contrast, when several models need to be compared, AIC weights are used — under the assumption that one of the models in the set is the optimal model. These models can be nested or non-nested (i.e., including alternative predictors). AIC weights therefore indicate the strength of the evidence in favor of a model in a particular set of competing models (Burnham & Anderson, 2003). The model with the highest AIC weight is taken as the one which best approximates the “true” process underlying the phenomenon under study, and the other models are evaluated in relation to that optimal model. This evaluation is based on Delta values, which correspond to the difference in AIC between the best model in the set and a particular competitor model. The following rules of thumb are usually applied to decide which models need to be considered along with the best model:

- (14) AIC weights interpretation (Burnham & Anderson, 2003)
- a. A model with a Delta value within 1-2 of the best model has substantial support in the data, and should be considered along with the best model.
 - b. A Delta value within only 4-7 units of the best model has considerably less support.
 - c. A Delta value > 10 indicates that the worse model has virtually no support and can be omitted from further consideration.

This method can be used for variable selection, i.e., to determine which predictor variable has the greatest influence, among a set of competitors (Burnham & Anderson, 2003). In our case, it can be used to compare the BPI with alternative language experience predictors. This is done by summing the Akaike weights of variables across all the models where the variables occur. The competing variables are then ranked using these sums. The larger this sum of weights, the more important the variable is.

Following the method explained above, we created a set of alternative models with the following alternative measures of language experience:

- (15) a. the BPI

- b. cumulative exposure the school language (English)
- c. current exposure to the school language (English)
- d. cumulative use the school language (English)
- e. current use of the school language (English)

As all the children in this sample were exposed to two languages only, school language measures and HL measures are equivalent from a statistical modelling (albeit yielding opposite coefficients). Here we report on the measures calibrated on the school language (English).

Using the optimal model for the bilingual children's Sentence Repetition data (reported in Table 7), we replaced the BPI with each of the above predictors in turn (i.e., refitting the model 4 times, once per predictor other than the BPI). As the models differed only with respect to the language experience measure, the one with the best fit can be considered the one with the most reliable predictor of language proficiency. The results of the comparison are summarized in Table 11, ordered by model weight.

There is a 37% probability that the model using Cumulative English Exposure as a predictor is the best in the set. When the difference in AIC between 2 models (Δ AIC or Delta) is < 2 , it is reasonably safe to consider that both models have approximately equal weight in the data (Burnham & Anderson, 2003). There is strong evidence (Delta = 1.30) for the need to take Cumulative English Use into account alongside Cumulative English Exposure — which is in effect what the BPI does. With a Delta value of 0.30, the BPI can be considered to be roughly as good a predictor of English language proficiency as Cumulative Exposure.¹⁹ It has a 32% likelihood of being the best model in the set — which is not far off the 37% likelihood of the model based on Cumulative English Exposure.

The answer to first part of Research Question 2 (as to whether a simpler measure of language experience might fare better than a complex measure) is that the BPI, as a combined measure of cumulative language exposure and use, is not more informative as a predictor of school language proficiency, compared with a simple measure of cumulative language exposure. As explained above, the BPI can be interpreted as a

measure of cumulative exposure to the HL, adjusted for HL use: two children with the same amount of Cumulative HL Exposure will have a different BPI if one of them is relatively passive in the HL (i.e., if s/he tends to answer back in English when addressed in the HL). However, the additional information encapsulated in the BPI (compared with a simple measure of cumulative language exposure) does not improve its predictive potential. This could be because language *use* does not have much of an effect over and above that of language *exposure*.

With respect to the cumulative-current comparison, cumulative measures are indeed more informative than current measures, as predictors of school language proficiency.

We now turn to the second part of Research Question 2, regarding the specific impact of language *use* as predictor of school language proficiency.

Question 2.2: Teasing apart the effect of exposure and use

If the amount of language use (estimated from the parental questionnaires) was perfectly correlated to the amount of language exposure, it would not be possible to disentangle their effect statistically as predictors of language proficiency. In the group under study, however, an imbalance between language use and exposure is observed: many children tended to be relatively “passive” in their home language (HL), as shown in Figure 10. Passivity reflects the extent to which the child used their HL in relation to the extent they were exposed to it, and it can be calculated by dividing Cumulative HL Use by Cumulative HL Exposure. Children with the highest score (100%) are as active in their home language as they are exposed to it. Lower scores indicate relative passivity in the home language: the child uses that language less than they are exposed to it (0% indicating no output in the home language at all, which was the case for 9 children). Importantly, a high score does not necessarily correlate with a high amount of HL exposure in our population sample: high usage is observed across the whole range of home language exposure.

————- FIG 10 HERE ————

Children who are relatively passive in their HL are (by implication) more active in their school language (henceforth SL), as they answer back in English in some or all of the contexts in which they are addressed in the HL. It is possible that this “extra” amount of English use confers a proficiency advantage in that language. This is what we investigate below.

As explained in the "Methods of analysis" section, it is not possible to include strongly correlated predictors in the same regression model, as is the case for our exposure and use measures. To remedy this problem, we derived a measure of the *difference* between Cumulative SL Exposure vs. use (by subtracting Cumulative SL Exposure from Cumulative SL Use). This new measure, which we will call SL Difference, is significantly but only marginally correlated with Cumulative SL Exposure ($r = -0.19$, $p < 0.0001$), which makes it possible to consider the two together in the same regression model.

We refitted the optimal model of the Sentence Repetition accuracy data (by item), with Cumulative SL Exposure and SL Difference as language experience predictors. The effect of SL Difference was not significant ($\beta = -0.005$, $p=0.84$), and allowing for an interaction between Cumulative SL Exposure and SL Difference led to non-convergence.

Given the lack of convergence with linear regression, we turned to a more powerful algorithm to investigate the relationship between SL exposure and use as predictors of SL proficiency.

The data was reanalyzed with a generalized additive mixed model (GAMM — Wood, 2006; Baayen, Vasishth, Kliegl, & Bates, 2017) using the R-package ‘mgcv’ (version 1.8-28). GAMMs are a variant of the linear models introduced above. Non-linear regression models operate according to the same broad principles as their linear variants, except that the regression line fitted through the scatterplot of observations can feature a certain amount of “wiggleness” instead of a straight line. The linear effects are included as *parametric terms* in the model (as above), and the non-linear effects are included as *smooths*.²⁰ The model has an in-built optimizer to avoid over-fitting by allowing too much “wiggleness” in the smooths. In the model

reported below, the random effects were modeled as smooths. To probe the relationship (or lack thereof) between SL cumulative exposure and SL Difference, we allowed a non-linear interaction of the two HL measures (Cumulative Exposure and HL Difference).

Allowing the non-linear interaction of Cumulative SL Exposure and SL Difference did yield a significant improvement in model fit, compared with a non-linear model without that interaction. The interaction between Cumulative SL Exposure and SL Difference reached a statistical significance ($\chi^2 = 10.45$, $\text{edf} = 3$, $p = 0.015$), suggesting a cross-over effect of exposure and use affecting mainly children with a high amount of cumulative exposure to the SL.

To test the robustness of the modest significance level of the interaction between language exposure and use, we fitted an alternative model allowing our other main environmental predictor of interest (i.e., SES) to interact non-linearly with language exposure.

This alternative model revealed a much more important non-linear interaction at play in the SRep data, which renders the impact of SL passivity non-significant (even in interaction with cumulative SL exposure): the non-linear interaction between socio-economic status and cumulative SL exposure turns out to be the strongest predictor of accuracy in the SRep test ($\chi^2 = 30.69$, $\text{edf} = 3$, $p = 0.000001$). Figure 11 shows that, at relatively low levels of cumulative exposure to the SL, SES does not confer an advantage in the SRep test; but at medium-to-high levels of cumulative SL exposure, accuracy in the SRep test is strongly correlated with SES.

————— FIG 11 HERE —————

Summary

In answer to Research Question 2, we have demonstrated that cumulative exposure to the school language is the best predictor of proficiency in that language, both compared with current measures, and compared with a combined measure of cumulative exposure and use. In attempting to tease apart the impact of language

exposure and language use, we discovered that language use was not a robust predictor of school language proficiency. The marginally significant non-linear interaction between language exposure and use was trumped by a very strong non-linear interaction between language exposure and SES. Children with below-average exposure to the school language did not benefit from an SES advantage (in terms of proficiency score). But in children with above-average exposure to the school language, performance improved exponentially as the levels of language exposure and SES increased.

We now turn to the “catch up” question (Research Question 3): from what amount of cumulative exposure to the school language do bilingual children perform within the monolingual range?

Question 3: Critical thresholds of language experience

Even within the same age group, monolingual children vary in terms of language proficiency (see Kidd, Donnelly, & Christiansen, 2018 for a recent review). Identifying what counts as monolingual-like performance needs to take that variability into account. One approach in the literature has been to rely on a threshold of 1.25 standard deviation below the mean to identify monolingual children at risk of a Developmental Language Disorder (Tomblin, Records, & Zhang, 1996; Conti-Ramsden, 2003). Any proficiency score above that threshold is considered “normal”, which suggests a substantial amount of variation within the “normal” range. Our purpose here is not to identify DLD among bilinguals, but to identify bilinguals who have caught up with their monolingual peers in the school language. One might suggest that this would require them to score above the -1.25 SD threshold (defined in relation to the mean achieved by their monolingual peers).

Instead of relying on an arbitrary threshold on raw scores (such as -1.25 SD), we propose to use an objective, data-driven method to identify the threshold in language experience beyond which bilingual children can be expected to perform within the monolingual range in terms of proficiency in the school language. This method will allow us to control for the other factors which we know have a significant impact on

language proficiency, such as SES and cognitive factors.

Below, we carry out the threshold analysis based on cumulative exposure to the school language (i.e., an additive measure) and on language dominance (i.e., a proportion measure) in turn, to address the role of cumulative exposure and language dominance.

Cumulative exposure thresholds

Our starting point is the optimal (non-linear, mixed-effect) model for the sentence repetition (SRep) data, which had been fitted to the target accuracy data for the bilingual children only. We rely on cumulative exposure as our reference measure for bilingual language experience, as this proved the best predictor in the population under study. We first refit this model to the entire cohort of children (including monolinguals) — using exposure to English as a predictor. When the model is fitted to the entire cohort using that predictor, age no longer accounts for any unique variance. We therefore exclude it from the model. After confirming the significance of the non-linear interaction between SES and Cumulative SL exposure in the whole-cohort model (not reported here), we perform the threshold analysis, as explained below.

The procedure for the threshold analysis consists in replacing the continuous SL exposure predictor with a binary one, for which the cutoff point would be determined objectively, based on Information Theory. To identify the optimal cutoff point, we re-fit the model recursively, using each time a different SL exposure value as the binary cut-off point. As the cumulative SL experience measure features 51 different values in the group under study, the model needs to be fitted 70 times. Plotting the AIC values for all the models (shown in Figure 12) shows the fluctuations in model fit depending on SL exposure threshold. The lowest AIC value identifies the model with the best fit. The cutoff point for that model is interpreted as the amount of exposure beyond which a bilingual child can be expected to perform within the monolingual range.

—————- FIG 12 HERE —————

The optimal cut-off point on the continuum of SL cumulative exposure is at 33

months equivalent (where the dashed line falls in Figure 12). The model therefore predicts that a (5- to 7-year-old) bilingual child who has been exposed to the school language for the equivalent at least 33 months can be expected to perform within the monolingual range (after controlling for SES and memory abilities) in the SRep task.

The same procedure is then repeated for the other aspects of English proficiency we assessed in this study.²¹ This results in the following thresholds of cumulative SL exposure: lexical semantics (composite DELV score): 44 months equivalent; discourse semantics: 32 months equivalent;— see Figure 13.

————— FIG 13 HERE —————

The overall picture, when combining the information from the three analyses above, is that from 32 or 33 months equivalent of cumulative exposure to the school language, bilingual children (between 5 and 7 years of age) perform within the range of monolingual children in *some* aspects of language proficiency (indexed by the lexical semantics and the sentence repetition tasks). The discourse semantics results suggest a more conservative threshold of 44 months equivalent (with respect to lexical semantics).

In the next section, we compare our threshold analysis to a more traditional approach based on language dominance.

Language dominance thresholds

In the literature on bilingualism, it is common practice to consider that, below 20% of current exposure to an additional language, children should be considered “functionally monolingual” in their dominant language (e.g., Bedore et al., 2012; Gutiérrez-Clellen & Kreiter, 2003; Pearson et al., 1997). As its name indicates, functional monolingualism implies that the child should perform within the monolingual range in terms of language proficiency.

Assuming that cumulative language exposure is a reliable proxy for language dominance (as demonstrated by Unsworth, 2015), we operationalized language dominance as the proportion of cumulative exposure to the school language out of the child’s lifetime (in months). We then replicated the threshold analyses above using that

estimate of language dominance as our predictor of interest.

The threshold analysis of the SRep data using language dominance as predictor of interest suggests a cut-off point of 57% dominance in the school language (in 5- to 7-year-olds).

Figure 14 compares the threshold based on the additive measure (corresponding to the horizontal line) with the threshold based on the proportional measure (cf. the vertical line). A number of data points are found in the top left quadrant, corresponding to the children that are predicted to perform within the monolingual range according to the additive threshold, but not according to the proportional threshold. The threshold analysis based on language dominance is therefore more conservative than that based on the cumulative measure of language exposure.

————— FIG 14 HERE —————

Figure 15 compares Bedore et al's categories with our objective threshold for monolingual-like performance.

————— FIG 15 HERE —————

Summary and discussion

The discussion is organized around our three research questions, which are repeated below for clarity.

Research Question 1: Do a gradient measure of bilingual language experience (combining exposure and use) and a gradient measure of socio-economic status (as a proxy for the richness of the child's language environment) significantly predict the following aspects of school language proficiency: comprehension and production of complex sentences, lexical semantics, and discourse-semantics? We proposed a gradient measure combining cumulative home language exposure and use (the Bilingualism Profile Index, BPI) and demonstrated that it significantly predicts 5- to 7-year old bilingual children's proficiency in the school language. This finding was replicated with respect to different aspects of English proficiency, including: morpho-syntax (indexed by the LITMUS sentence repetition test), lexical semantics (indexed by the DELV

lexical tests) and discourse-semantics (indexed by the DELF articles test). Our results are in line with a substantial body of research demonstrating the impact of language exposure on language proficiency in bilinguals (see Paradis, 2017 for a recent review), and expand on the bulk of these studies by considering a range of proficiency measures.

Some aspects of language were found to be more sensitive than others to language exposure. Functional errors in the sentence repetition task were predicted by language experience more strongly than lexical errors, while inflectional errors were not predicted by language experience at all, suggesting that a sufficient amount had already been received by most children in this study to acquire those aspects of English grammar. This is in line with studies showing that the non-uniform impact of language exposure on different aspects of grammar reflects their acquirability (Tsimpli, 2014; Schulz & Grimm, 2019).

Children's socio-economic status was a significant source of substantial individual differences, over and above the effect of the cognitive predictors (which are independently known to correlated with SES — see e.g., Gathercole et al., 2016). Along with many others (e.g., Bohman et al., 2010; Hoff, 2013; Meir & Armon-Lotem, 2017), we interpret SES as a proxy for the richness of the (linguistic) environment experienced by the child, which results in qualitative and quantitative differences in children's language experience. Further research will be required to identify which of the correlates of SES genuinely explains the nature of its effect. One potentially important factor that we did not consider in the present study is the home literacy environment and the extent to which reading and writing in the home language and/or the school language correlates with SES and in turn affects children's language skills. Brinchmann, Braeken, and Lyster (2019), for instance, argue that the home literacy environment accounts for individual differences in vocabulary and grammar in preschool monolingual children. A hypothesis for future research is whether (and to what extent) the amount and diversity of literacy activities explain the relationship between SES and language proficiency in bilingual children.

SES was found to predict children's performance in the sentence repetition task

and in the discourse semantics tasks, but not in the lexical semantics tasks. We speculate that this could be due to the very strong impact of cognitive predictors on the latter measures. SES is a strong predictor of performance on cognitive flexibility (as demonstrated by De Cat, Gusnanto, & Serratrice, 2018 for this group of children), and might therefore have only an indirect impact on performance in the lexical semantics task.

An important finding of this study is that SES interacts with language experience in a complex way, which we modeled as the non-linear interaction of SES and cumulative exposure to the school language: the extent to which SES confers an advantage in terms of school language proficiency depends on the amount exposure to that language. In our sample, children with below-average cumulative exposure to the school language hardly benefited from any SES advantage, whereas at higher levels of cumulative exposure to the school language, that advantage grew exponentially in relation to school language exposure. In other words, SES was only associated with a higher proficiency score in the children who had received a substantial amount of exposure to the school language over their lifetime. What our cross-sectional study was not able to investigate is the possibility that children from higher SES might be able to catch up faster with their monolingual peers. This will need to await future research based on longitudinal data.

Research question 2 sought to identify the optimal way of modeling language experience, as a predictor of school language proficiency. We investigated it in three steps.

How does a combined measure of language experience (combining exposure and use) compare with simpler measures (i.e., exposure *or* use), as predictors of school language proficiency? Adopting an information-theoretic approach, we found that a more precise measure of cumulative language experience (the Bilingualism Profile Index, which encompasses exposure and use) did not improve predictive accuracy compared with a simpler measure of cumulative HL exposure. This finding will be discussed in relation to the issue of language exposure vs. use below.

Are *cumulative* measures more accurate than current measures, in predicting school language proficiency? To our knowledge this is the first study to systematically compare different measures of language experience as predictors of language proficiency in bilingual children (i.e., current exposure, current use, cumulative exposure, cumulative use, and a combined measure of cumulative exposure and use). Our information-theoretic comparison demonstrated that cumulative experience measures fare better as predictors of language proficiency than current experience measures. Future research will be needed to determine if this is the case in older age groups as well.

What is the specific impact of language *use* as a predictor of school language proficiency? Analyses using decorrelated measures confirmed that there was no robust evidence for an effect of language use over and above the effect of exposure to the school language. In other words, if a child uses the school language even in (some of) the contexts in which they are addressed in the home language, that “extra” use does not confer a tangible advantage in terms of school language proficiency. This is likely due to the leveling effect of schooling, and/or possibly to the fact that the children’s English utterances in HL contexts are not elaborated upon in English by their interlocutor. We hypothesize that language use will however be a significant predictor of proficiency in the home language (as many children were relatively “passive” in that language). The Bilingualism Profile Index might thus prove to be more accurate in predicting *home* language proficiency. This could not be investigated in this study, given the number of home languages represented in our sample.

Research question 3 focused on the identification of functional thresholds in bilingual language experience in relation to school language proficiency: What is the critical amount of school language experience required for bilingual children to perform within the monolingual range? In other words: at what point can one expect the gap between bilinguals and monolinguals to be “closed” in the school language? We exploited an objective, data-driven method to investigate this based on (a) an additive measure of language experience, and (b) a language dominance measure.

Once socio-economic status and cognitive abilities have been taken into account,

the threshold of exposure required for bilingual children to perform within the monolingual range was found to vary depending on the proficiency measure. The measures of discourse semantics (indexed by the mastery of definiteness distinctions) and complex morpho-syntax (indexed by the sentence repetition task) showed that children performed within the monolingual range from 32 or 33 months-equivalent of exposure to the school language. A more conservative threshold of 44 months-equivalent was suggested by the lexical semantics measure. It is not surprising to find that different aspects of language proficiency show different levels of sensitivity to language exposure, as already alluded to above. R. Jia and Paradis (2017) show that morpho-syntactic structures encoding whether a referent is new or old information can be acquired by bilingual children even with limited language exposure. This is consistent with our finding that the lowest amount of exposure to English required for performance within the monolingual range was observed in the discourse-semantics task (which tested children's knowledge of the *a / the* distinction in English). However, the higher threshold observed with respect to the lexical semantics measure needs to be interpreted with caution. That test proved the most challenging both for bilinguals and monolinguals (with a mean overall score of 63% in the monolinguals, compared with 81% in the sentence repetition test, 83% in the sentence comprehension test and 77% in the discourse-semantics test) and it is possible that the higher threshold might in fact be due to age as a latent variable (which in turn could be due in part to the cognitive demands of the task).

If expressed in terms of language dominance (rather than cumulative experience), our results indicate that, in this age group, children with at least 57% of exposure to the school language (averaged over their lifetime) are likely to perform within the monolingual range. This aligns closely with the 60% threshold suggested for toddlers by Cattani et al. (2014), but is more lenient than a 80% language exposure cut-off point commonly used for "functional monolingualism".

The approach adopted by Bedore et al. (2012) is more lenient too, but for a different reason. For them, functional monolingualism obtains if there is at least 80%

exposure to *or* use of a language (and 20% or less exposure or use of another language). This disjunctive criterion gives equal weight to exposure and use: if either falls below 20%, the child is categorized as functionally monolingual in their other language. According to the criteria proposed by Bedore et al. (2012), 41 of the 87 bilinguals in our sample would be classified as functionally monolingual in English. Yet, many children in that group perform below the monolingual range, as identified by our threshold analyses (see Figure 15).

One reason for this discrepancy could be that Bedore et al.'s classification of language dominance is based on current language experience, which we have seen is less informative a predictor of language proficiency than cumulative language experience in our sample. Another, likely more important, reason is that their classification assigns equal weight to language exposure and language use, as explained above. While we agree it is important to take language use into account, it appears not to have much impact on proficiency *in the school language*, as pointed above. Further research will be required to determine objectively the relative importance of language exposure and use as components of bilingual language experience. For instance, language use is likely to have a significant impact on proficiency in the home language, and possibly on executive functions in the cognitive domain. The investigation of phenomena that are sensitive to language use (over and above language experience) will provide fertile grounds to evaluate how it should be modeled. Should language use and exposure have equal weight, as proposed by Bedore et al. (2012)? Or should it have a weaker impact, as modeled in our Bilingualism Profile Index? We leave it to future research to evaluate the merit of each approach, including whether the language of the conversational partner has an impact.

Finally, three cautionary notes are in order, regarding the interpretation of our findings.

First, while critical thresholds can be useful for practical purposes (e.g. selection criteria for experimental studies, or the identification of children in need of support with their school language), we agree with Treffers-Daller (2015), Luk (2015) and many

others that language dominance and bilingualism are best treated as continua rather than discrete categories.

Second, the functional thresholds identified need to be interpreted in light of other significant predictors, i.e., SES and cognitive factors. For instance, a child with more than 60% exposure to the school language over her lifetime will not necessarily score within the monolingual range of proficiency if she comes from a deprived background. All the proficiency tests used in this study required a certain amount of cognitive effort, and this induced a significant amount of individual variation in the scores. Short term memory had a significant impact on performance in the sentence repetition test²² and the lexical-semantic tests. Working memory had a significant impact on performance in the lexical semantic tasks, but not the sentence repetition task nor the discourse semantic task. The Sentence Structure test of the CELF proved to be more sensitive to children's cognitive abilities than to their mastery of complex syntax, due to an overwhelming confound caused by the cognitive demands of the task (see De Cat, under review). It was therefore not discussed in the present paper.

Third, the children in this study were all in the early stages of formal education, and most bilinguals will have caught up with their monolingual peers by the end of primary school — except if part of an at-risk group (which we were not able to investigate here; see Strand & Demie, 2005). Although bilingualism might result in an apparent delay in terms of school language proficiency in the early years of formal education, it tends to turn into an asset once the proficiency gap between monolinguals and bilinguals has narrowed sufficiently (or closed). If bilingual children are compared with monolinguals of similar levels of English proficiency, the bilingual group already shows an advantage in social, emotional and behavioral functioning in Reception year (around 5 years of age), and a functional advantage in meeting curriculum targets at Year 2 (around 7 years of age) — see e.g., Dowdy, Dever, DiStefano, and Chin (2011); Halle, Hair, Wandner, McNamara, and Chien (2012); Whiteside, Gooch, and Norbury (2017).

Investigating language development in its linguistic and cognitive complexity in a

heterogeneous population of children presents non-trivial challenges. The analytical approach taken here goes some way to propose a solution to dealing with complex data and to making sense of the multi-faceted bilingual experience that is common in many schools today.

References

- Avons, S. E., Wragg, C. A., Cupples Wragg, L., & Lovegrove, W. J. (1998). Measures of phonological short-term memory and their relationship to vocabulary development. *Applied Psycholinguistics*, *19*, 583-601. doi: 10.1017/S0142716400010377
- Baayen, R. H. (2008). *Analyzing linguistic data. A practical introduction to statistics using R*. Cambridge: CUP.
- Baayen, R. H., Vasishth, S., Kliegl, R., & Bates, D. (2017). The cave of shadows. Addressing the human factor with generalized additive mixed models. *Journal of Memory and Language*, *94*, 206-234. doi: 10.1016/j.jml.2016.11.006
- Baddeley, A. (2000). The episodic buffer: A new component of working memory? *Trends Cognitive Science*, *4*, 417-423. doi: 10.1016/S1364-6613(00)01538-2
- Baddeley, A., Gathercole, S. E., & Papagno, C. (1998). The phonological loop as a language learning device. *Psychological Review*, *105*, 158-173. doi: 10.1016/S1364-6613(00)01538-2
- Barac, R., & Bialystok, E. (2012). Bilingual effects on cognitive and linguistic development: Role of language, cultural background, and education. *Child Development*, *83*, 413-422. doi: 10.1111/j.1467-8624.2011.01707.x
- Bedore, L., Peña, E., Summers, C., Boerger, K., Resendiz, M., Greene, K., ... Gillam, R. (2012). The measure matters: Language dominance profiles across measures in Spanish-English bilingual children. *Bilingualism: Language and Cognition*, *15*, 616-629. doi: 10.1017/S1366728912000090
- Bernardini, P., & Schlyter, S. (2004). Growing syntactic structure and code-mixing in the weaker language: The Ivy hypothesis. *Bilingualism: Language and Cognition*, *7*, 49-69. doi: 10.1017/S1366728904001270
- Bohman, T. M., Bedore, L., Peña, E. D., Mendez-Perez, A., & Gillam, R. B. (2010). What you hear and what you say: Language performance in Spanish-English bilinguals. *International Journal of Bilingual Education and Bilingualism*, *13*, 325-344. doi: 10.1080/13670050903342019

- Boyle, W., Lindell, A. K., & Kidd, E. (2013). Investigating the role of verbal working memory in young children's sentence comprehension. *Language Learning, 63*, 211-242. doi: 10.1111/lang.12003
- Brinchmann, E. I., Braeken, J., & Lyster, S.-A. H. (2019). Is there a direct relation between the development of vocabulary and grammar? *Developmental Science, 22*, e12709. doi: 10.1111/desc.12709
- Burnham, K., & Anderson, D. (2003). *Model selection and multimodel inference: A practical information-theoretic approach*. New York: Springer Science & Business Media.
- Calvo, A., & Bialystok, E. (2014). Independent effects of bilingualism and socioeconomic status on language ability and executive functioning. *Cognition, 130*, 278-288. doi: 10.1016/j.cognition.2013.11.015
- Carroll, S. E. (2017). Exposure and input in bilingual development. *Bilingualism: Language and Cognition, 20*, 3-16. doi: 10.1017/S1366728915000863
- Cattani, A., Abbot-Smith, K., Farag, R., Krott, A., Arreckx, F., Dennis, I., & Floccia, C. (2014). How much exposure to English is necessary for a bilingual toddler to perform like a monolingual peer in language tests? *International Journal of Language & Communication Disorders, 49*, 649-671. doi: 10.1111/1460-6984.12082
- Chiat, S., & Polišenská, K. (2016). A framework for crosslinguistic nonword repetition tests: Effects of bilingualism and socioeconomic status on children's performance. *Journal of Speech, Language, and Hearing Research, 59*, 1179-1189. doi: 10.1044/2016_JSLHR-L-15-0293
- Conti-Ramsden, G. (2003). Processing and linguistic markers in young children with specific language impairment (SLI). *Journal of Speech, Language, and Hearing Research, 46*, 1029-1037. doi: 10.1044/1092-4388(2003/082)
- Cook, V. (1994). Universal grammar and the learning and teaching of second languages. In T. Odlin (Ed.), *Perspectives on pedagogical grammar* (p. 25-48). Cambridge: CUP.

- De Cat, C. (under review). What does the CELF Sentence Structure test index in monolingual and bilingual children?
- De Cat, C., Gusnanto, A., & Serratrice, L. (2018). Identifying a threshold for the executive function advantage in bilingual children. *Studies in Second Language Acquisition*, *40*, 119-151. doi: 10.1017/S0272263116000486
- De Houwer, A. (2007). Parental language input patterns and children's bilingual use. *Applied Psycholinguistics*, *28*, 411-424. doi: 10.1017/S0142716407070221
- De Houwer, A. (2009). *Bilingual first language acquisition*. Clevedon: Multilingual Matters.
- De Houwer, A. (2011). Language input environments and language development in bilingual acquisition. *Applied Linguistics Review*, *2*, 221-240. doi: 10.1515/9783110239331.221
- Dowdy, E., Dever, B. V., DiStefano, C., & Chin, J. K. (2011). Screening for emotional and behavioral risk among students with limited english proficiency. *School Psychology Quarterly*, *26*, 14.
- Engel de Abreu, P., Baldassi, M., Puglisi, M., & Bepi-Lopes, D. (2013). Cross-linguistic and cross-cultural effects on verbal working memory and vocabulary: Testing language-minority children with an immigrant background. *Journal of Speech, Language and Hearing Disorders*, *56*, 630-42. doi: 10.1044/1092-4388(2012/12-0079)
- Engel de Abreu, P., Gathercole, S., & Martin, R. (2011). Disentangling the relationship between working memory and language: The roles of short-term storage and cognitive control. *Learning and Individual Differences*, *21*, 569-574. doi: 10.1016/j.lindif.2011.06.002
- Floccia, C., Sambrook, T. D., Delle Luche, C., Kwok, R., Goslin, J., White, L., . . . Plunkett, K. (2018). *Vocabulary of 2-year-olds learning English and an Additional Language: Norms and effects of linguistic distance*. Wiley.
- Friedmann, N., & Rusou, D. (2015). Critical period for first language: The crucial role of language input during the first year of life. *Current Opinion in Neurobiology*,

35, 27-34. doi: 10.1016/j.conb.2015.06.003

Gathercole, V. C. M., Kennedy, I., & Môt Thomas, E. (2016). Socioeconomic level and bilinguals' performance on language and cognitive measures. *Bilingualism: Language and Cognition*, 19, 1057-1078. doi: 10.1017/S1366728915000504

Genesee, F., Nicoladis, E., & Paradis, J. (1995). Language differentiation in early

bilingual development. *Journal of Child Language*, 22, 611-632. doi:

10.1017/S0305000900009971

Goldstein, B., Bunta, F., Lange, J., Rodriguez, J., & Burrows, L. (2010). The effects of measures of language experience and language ability on segmental accuracy in

bilingual children. *American Journal of Speech-Language Pathology*, 19, 238-247.

doi: 10.1044/1058-0360(2010/08-0086)

Goldthorpe, J. (1980). *Social mobility and class structure in modern Britain*. Oxford: Clarendon.

Grosjean, F. (1989). Neurolinguists, beware! the bilingual is not two monolinguals in one person. *Brain and Language*, 36, 3-15. doi: 10.1016/0093-934X(89)90048-5

Grosjean, F. (2016). The complementarity principle and its impact on processing, acquisition, and dominance. In C. Silva-Corvalan & J. Treffers-Daller (Eds.), *Language dominance in bilinguals: Issues of operationalization and measurement* (p. 66-84). Cambridge: CUP.

Grüter, T., & Paradis, J. (Eds.). (2014). *Input and experience in bilingual development*. Amsterdam: John Benjamins.

Gutiérrez-Clellen, V. F., & Kreiter, J. (2003). Understanding child bilingual acquisition using parent and teacher reports. *Applied Psycholinguistics*, 24, 267-288. doi:

10.1017/S0142716403000158

Halle, T., Hair, E., Wandner, L., McNamara, M., & Chien, N. (2012). Predictors and outcomes of early vs. later English language proficiency among English language

learners. *Early childhood research quarterly*, 27, 1-20. doi:

10.1016/j.ecresq.2011.07.004

Hoff, E. (2006). How social contexts support and shape language development.

- Developmental Review*, 26, 55-88. doi: 10.1016/j.dr.2005.11.002
- Hoff, E. (2013). Interpreting the early language trajectories of children from low-SES and language minority homes: Implications for closing achievement gaps. *Developmental Psychology*, 49, 4. doi: 10.1037/a0027238
- Hoff, E., Core, C., Place, S., Rumiche, R., Señor, M., & Parra, M. (2012). Dual language exposure and early bilingual development. *Journal of Child Language*, 39, 1-27. doi: 10.1016/j.cogpsych.2010.08.002
- Huttenlocher, J., Waterfall, H., Vasilyeva, M., & Vevea, J. (2010). Sources of variability in children's language growth. *Cognitive Psychology*, 16, 343-365.
- Jardak, A., & Byers-Heinlein, K. (2018). Labels or concepts? The development of semantic networks in bilingual two-year-olds. *Child Development*, 90, e212-e229. doi: 10.1111/cdev.13050
- Jia, G., & Aaronson, D. (2003). A longitudinal study of Chinese children and adolescents learning English in the United States. *Applied Psycholinguistics*, 24, 131-161. doi: 10.1017/S0142716403000079
- Jia, G., Aaronson, D., & Wu, Y. (2002). Long-term language attainment of bilingual immigrants: Predictive variables and language group differences. *Applied Psycholinguistics*, 23, 599-621. doi: 10.1017/S0142716402004058
- Jia, R., & Paradis, J. (2017). The use of referring expressions in narratives by Mandarin heritage language children and the role of language environment factors in predicting individual differences. *Bilingualism: Language and Cognition*, 18, 737-752. doi: 10.1017/S1366728914000728
- Kaushanskaya, M., & Prior, A. (2015). Variability in the effects of bilingualism on cognition: It is not just about cognition, it is also about bilingualism. *Bilingualism: Language and Cognition*, 18, 27-28. doi: 10.1017/S1366728914000510
- Kidd, E. (2013). The role of verbal working memory in children's sentence comprehension. *Topics in Language Disorders*, 33, 208-223. doi: 10.1097/TLD.0b013e31829d623e

- Kidd, E., Donnelly, S., & Christiansen, M. H. (2018). Individual differences in language acquisition and processing. *Trends in Cognitive Sciences*, 22, 154-169. doi: 10.1016/j.tics.2017.11.006
- Kupisch, T. (2007). Determiners in bilingual german-italian children: What they tell us about the relation between language influence and language dominance. *Bilingualism: Language and Cognition*, 10, 57-78. doi: 10.1017/S1366728906002823
- Laloi, A., de Jong, J., & Baker, A. (2017). Can executive functioning contribute to the diagnosis of sli in bilingual children? *Linguistic Approaches to Bilingualism*, 7, 431-459. doi: 10.1075/lab.15020.lal
- Lanza, E. (2004). *Language mixing in infant bilingualism: A sociolinguistic perspective*. Oxford: Clarendon Press.
- Lauro, L. J. R., Reis, J., Cohen, L. G., Cecchetto, C., & Papagno, C. (2010). A case for the involvement of phonological loop in sentence comprehension. *Neuropsychologia*, 48, 4003-4011. doi: 10.1016/j.neuropsychologia.2010.10.019
- Luk, G. (2015). Who are the bilinguals (and monolinguals)? *Bilingualism: Language and Cognition*, 18, 35-36. doi: 10.1017/S1366728914000625
- Magimairaj, B., & Montgomery, J. (2012). Children's verbal working memory: Role of processing complexity in predicting spoken sentence comprehension. *Journal of Speech, Language and Hearing Disorders*, 55, 669-682. doi: 10.1044/1092-4388(2011/11-0111)
- Majerus, S., Poncelet, M., Greffe, C., & Van der Linden, M. (2006). Relations between vocabulary development and verbal short-term memory: The relative importance of short-term memory for serial order and item information. *Journal of experimental child psychology*, 93, 95-119. doi: 10.1016/j.jecp.2005.07.005
- Marinis, T., & Armon-Lotem, S. (2015). Sentence repetition. In S. Armon-Lotem, J. de Jong, & N. Meir (Eds.), *Assessing multilingual children. Disentangling bilingualism from language impairment* (p. 95-122). Bristol: Multilingual Matters.
- Marinis, T., Chiat, S., Armon-Lotem, S., Gibbons, D., & Gipps, E. (2010). *School-Age*

- Sentence Imitation Test (SASIT)*. Reading: University of Reading.
- McDonald, J. (2008). Grammaticality judgments in children: The role of age, working memory and phonological ability. *Journal of Child Language*, *35*, 247-268. doi: 10.1017/S0305000907008367
- Meir, N., & Armon-Lotem, S. (2017). Independent and combined effects of socioeconomic status (SES) and bilingualism on children's vocabulary and verbal short-term memory. *Frontiers in Psychology*, *8*, 1442. doi: 10.3389/fpsyg.2017.01442
- Meisel, J. M. (2009). Second language acquisition in early childhood. *Zeitschrift für Sprachwissenschaft*, *28*, 5-34. doi: 10.1515/ZFSW.2009.002
- Montgomery, J. (1995). Sentence comprehension in children with specific language impairment: The role of phonological working memory. *Journal of Speech and Hearing Research*, *38*, 187-199. doi: 10.1044/jshr.3801.187
- Montgomery, J., Magimairaj, B., & O'Malley, M. (2008). Role of working memory in typically developing children's complex sentence comprehension. *Journal of Psycholinguistic Research*, *37*, 331-354. doi: 10.1007/s10936-008-9077-z
- Montrul, S. (2015). Dominance and proficiency in early and late bilingualism. In C. Silva-Corvalán & J. Treffers-Daller (Eds.), *Language dominance in bilinguals: Issues of measurement and operationalization* (p. 15-35). Cambridge: CUP.
- Papagno, C., Cecchetto, C., Reati, F., & Bello, L. (2007). Processing of syntactically complex sentences relies on verbal short-term memory: Evidence from a short-term memory patient. *Cognitive Neuropsychology*, *24*, 292-311. doi: 10.1080/02643290701211928
- Paradis, J. (2011). Individual differences in child English second language acquisition: Comparing child-internal and child-external factors. *Linguistic Approaches to Bilingualism*, *1*, 213-237. doi: 10.1075/lab.1.3.01par
- Paradis, J. (2017). Parent report data on input and experience reliably predict bilingual development and this is not trivial. *Bilingualism, Language and Cognition*, *27*-28. doi: 10.1017/S136672891600033X

- Paradis, J., Crago, M., Genesee, F., & Rice, M. (2003). French-English bilingual children with SLI: How do they compare with their monolingual peers? *Journal of Speech, Language, and Hearing Research, 46*, 113-127. doi: 10.1044/1092-4388(2003/009)
- Paradis, J., Rusk, B., Duncan, T. S., & Govindarajan, K. (2017). Children's second language acquisition of English complex syntax: The role of age, input, and cognitive factors. *Annual Review of Applied Linguistics, 37*, 148-167. doi: 10.1017/S0267190517000022
- Pearson, B., Fernández, S. C., Lewedeg, V., & Oller, D. K. (1997). The relation of input factors to lexical learning by bilingual infants. *Applied Psycholinguistics, 18*, 41-58. doi: 10.1017/S0142716400009863
- Place, S., & Hoff, E. (2011). Properties of dual language exposure that influence two-year-olds' bilingual proficiency. *Child Development, 82*, 1834-1849. doi: 10.1111/j.1467-8624.2011.01660.x
- Ribot, K., Hoff, E., & Burrige, A. (2017). Language use contributes to expressive language growth: Evidence from bilingual children. *Child Development, 89*, 929-940. doi: 10.1111/cdev.12770
- Romaine, S. (1989). *Bilingualism*. Oxford: Blackwell.
- Rowe, M. L. (2012). A longitudinal investigation of the role of quantity and quality of child-directed speech in vocabulary development. *Child Development, 83*, 1762-1774. (1467-8624) doi: 10.1111/j.1467-8624.2012.01805.x
- Schafer, R., & de Villiers, J. (2000). Imagining articles: What *a* and *the* can tell us about the emergence of DP. In S. C. Howell, S. Fish, & T. Keith-Lucas (Eds.), *Proceedings of BUCLD 24* (p. 609-620). Somerville, MA: Cascadilla Press.
- Schulz, P., & Grimm, A. (2019). The age factor revisited: Timing in acquisition interacts with age of onset in bilingual acquisition. *Frontiers in Psychology, 9*, 2732. doi: 10.3389/fpsyg.2018.02732
- Semel, E., Wiig, E., & Secord, W. (2006). *Clinical Evaluation of Language Fundamentals* (Fourth ed.). London: Harcourt Assessment.

- Seymour, H. N., Roeper, T., & de Villiers, J. (2005). *DELV-NR (Diagnostic Evaluation of Language Variation) Norm-Referenced Test*. San Antonio TX: The Psychological Corporation.
- Silva-Corvalán, C., & Treffers-Daller, J. (Eds.). (2015). *Language dominance in bilinguals: Issues of measurement and operationalization*. Cambridge: CUP.
- Strand, S., & Demie, F. (2005). English language acquisition and educational attainment at the end of primary school. *Educational Studies*, *31*, 275-291. doi: 10.1080/03055690500236613
- Thordardottir, E. (2011). The relationship between bilingual exposure and vocabulary development. *International Journal of Bilingualism*, *15*, 426-445.
- Thordardottir, E., & Brandeker, M. (2013). The effect of bilingual exposure versus language impairment on nonword repetition and sentence imitation scores. *Journal of Communication Disorders*, *46*, 1-16. doi: 10.1016/j.jcomdis.2012.08.002
- Thordardottir, E., Rothenberg, A., Rivard, M.-E., & Naves, R. (2006). Bilingual assessment: Can overall proficiency be estimated from separate assessment of two languages? *Journal of Multilingual Communication Disorders*, *4*, 1-21. doi: 10.1080/14769670500215647
- Thorn, A., & Gathercole, S. (1999). Language-specific knowledge and short-term memory in bilingual and non-bilingual children. *The Quarterly Journal of Experimental Psychology Section A*, *52*, 303-324. doi: 10.1080/713755823
- Tomblin, J. B., Records, N. L., & Zhang, X. (1996). A system for the diagnosis of specific language impairment in kindergarten children. *Journal of Speech, Language, and Hearing Research*, *39*, 1284-1294. doi: 10.1044/jshr.3906.1284
- Treffers-Daller, J. (2015). Language dominance: The construct, its measurement, and operationalization. In C. Silva-Corvalán & J. Treffers-Daller (Eds.), *Language dominance in bilinguals: Issues of measurement and operationalization* (p. 235-265). Cambridge: CUP.
- Tsimpli, I. M. (2014). Early, late or very late?: Timing acquisition and bilingualism. *Linguistic Approaches to Bilingualism*, *4*, 283-313. doi: 10.1075/lab.4.3.01tsi

- Unsworth, S. (2013a). Assessing age of onset effects in (early) child L2 acquisition. *Language Acquisition, 20*, 74-92. doi: 10.1080/10489223.2013.766739
- Unsworth, S. (2013b). Assessing the role of current and cumulative exposure in simultaneous bilingual acquisition: The case of Dutch gender. *Bilingualism: Language and Cognition, 16*, 86-110. doi: 10.1017/S1366728912000284
- Unsworth, S. (2015). Amount of exposure as a proxy for dominance in bilingual language acquisition. In C. Silva-Corvalán & J. Treffers-Daller (Eds.), *Language dominance in bilinguals: Issues of measurement and operationalization*. Cambridge: CUP.
- Unsworth, S. (2017). Quantity and quality of language input in bilingual language development. In E. Nicoladis & S. Montanari (Eds.), *Lifespan perspectives on bilingualism* (p. 136-196). Berlin: Mouton de Gruyter/APA.
- Unsworth, S., Persson, L., Prins, T., & De Bot, K. (2014). An investigation of factors affecting early foreign language learning in the Netherlands. *Applied Linguistics, 36*, 527-548. doi: 10.1093/applin/amt052
- Verhagen, J., & Leseman, P. (2016). How do verbal short-term memory and working memory relate to the acquisition of vocabulary and grammar? A comparison between first and second language learners. *Journal of Experimental Child Psychology, 141*, 65-82.
- Wechsler, D. (1991). *WISC-III: Wechsler Intelligence Scale for Children: Manual*. San Antonio, TX: Psychological Corporation, Harcourt Brace Jovanovich.
- Whiteside, K. E., Gooch, D., & Norbury, C. F. (2017). English language proficiency and early school attainment among children learning English as an Additional Language. *Child Development, 88*, 812-827. doi: 10.1111/cdev.12615
- Willis, C. S., & Gathercole, S. E. (2001). Phonological short-term memory contributions to sentence processing in young children. *Memory, 9*, 349-363. doi: 10.1080/09658210143000155
- Wood, S. (2006). *Generalised additive models: An introduction with R*. Boca Raton, FL: Chapman and Hall/CRC.

Zelazo, P. (2006). The Dimensional Change Card Sort (DCCS): A method of assessing executive function in children. *Nature Protocols*, 1, 297-301. doi:
10.1038/nprot.2006.46

Footnotes

¹Percentages are given for those languages representing more than 5% of the sample.

²The Goldthorpe class scheme allocates individuals and families into categories of social class, based on their income, degree of economic security and chances of economic advancement, as well as on their degree of autonomy in performing their work-tasks and roles.

³Parents were asked “How well do you speak English?”, and could choose between “not at all”, “not well”, “quite well”, and “very well”.

⁴If there was more than one interlocutor in a time window, the time window was divided by the number of interlocutors.

⁵The version we used was called at the time the *School-Age Sentence Imitation Test*.

⁶After three questions, the prompt sentence is repeated to help the child remember it.

⁷We refer the interested reader to (De Cat et al., 2018) for the analysis of the cognitive measures used in this study.

⁸The fit of a model estimates how closely it matches the observed values in the dataset. The Akaike Information Criterion is an estimate of the model fit, penalized for over-fitting (i.e., the inclusion of too many parameters). The smaller the AIC, the better the model.

⁹The R-squared of a model expresses how much variance is captured by the model.

¹⁰Standardized scores, also known as z-scores, are rescaled so that they have a mean of zero and a standard deviation of one. Standardized scores were derived for all children as one group.

¹¹Current HL measures: $r = 0.86, p < .0001$; cumulative HL measures: $r = 0.93, p < .0001$.

¹²This Principal Component Analysis did **not** include any other variable than the cumulative measures of HL exposure and use.

¹³Loadings are the linear combinations of coefficients that correspond to the covariances/correlations between the original variables and the unit-scaled components. They are used to interpret principal components. We reversed the scores for ease of interpretability, so that a high value corresponds to a high amount of HL experience.

¹⁴The BPI correlates significantly with current exposure (Pearson’s product-moment correlation $r = .93, p < 0.0001$), current use (Pearson’s product-moment correlation $r = .89, p < 0.0001$), cumulative exposure (Pearson’s product-moment correlation $r = .99, p < 0.0001$) and cumulative use (Pearson’s product-moment correlation $r = .99, p < 0.0001$) in the HL.

¹⁵Dispersion is relatively modest in the case of cumulative measures. We come back to this in the comparative analyses below.

¹⁶As seen in Figure 4, performance according to the Full Accuracy analysis appears much poorer in both bilingual and monolingual children, with bilingual children only achieving a mean score of 35% (compared with 55% for the monolinguals) and several bilingual children obtaining an overall score of

zero.

¹⁷The estimates are expressed as the log of the odd ratio.

¹⁸Whether sequential bilingualism also affects language development qualitatively remains a controversial issue — see e.g., (Meisel, 2009; Unsworth, 2013a). We do not address it in this study.

¹⁹The results for HL measures (not reported here) are exactly identical to their English counterpart reported in Table 11, so the fact that the BPI was based on HL measures does not make a difference in this respect.

²⁰A smooth is a function that determines the wiggleness of the line (or surface, in the case of interactions).

²¹This was not done for the CELF Sentence Structure test, as language experience was not found to be a significant predictor of performance in the bilinguals.

²²This was entirely expected, as this test can be used as a verbal measure of working memory. In our study we scored children's performance based on the correct repetition of the target structure rather than exact word-for-word repetition. This alleviated the impact of memory factors.

	Gender	Min.	Max	Mean	St.Dev.
Bilinguals	F (n = 44)	5;1	6;9	5;10	0;5
(n = 87)	M (n = 43)	5;1	7;0	5;10	0;6
Monolinguals	F (n = 52)	5;0	7;0	6;0	0;7
(n = 87)	M (n = 35)	5;0	7;0	6;0	0;7

Table 1

Participant distribution in gender and age (in months)

	0	1	2	3	4	5
HL used with all	13	7	5	8	1	0
HL used with parents only	7	3	6	8	4	1
One-parent one-language	21	0	1	2	0	0

Table 2

Distribution of children according to onset of bilingual exposure (in years), by home language policy categories

Level	Structure	Example
1	Declarative with one auxiliary	The boy must sweep the floor in the kitchen.
	Short actional passive	The children were taken to the office.
	Object wh-question with <i>what/who</i>	What did the princess buy last month?
2	Declarative with two auxiliaries	The kitten could have hit the ball down the stairs.
	Long actional passive	She was seen by the doctor in the morning.
	Object wh-question with <i>which</i>	Which picture did he paint at home yesterday?
	Temporal clause	He will feed the cow before he waters the plants.
3	Conditional clause	If the kids behave we will go into the garden.
	Object relative clause	The bee that the man swallowed had hurt him.

Table 3

Sentence structures by level of difficulty in the SRep test

	3	4	5	6	7
Monolinguals	9	34	35	8	1
Bilinguals	13	34	37	2	1

Table 4

Forward Digit Recall (number of digits correctly recalled)

	0	2	3	4
Monolinguals	0	32	52	3
Bilinguals	3	34	46	4

Table 5

Backward Digit Recall (number of digits correctly reversed)

DCCS score	0	1	2	3
Monolinguals	0	2	26	59
	0	1%	15%	34%
Bilinguals	1	12	39	35
	1%	7%	22%	20%

Table 6

Distribution of overall DCCS scores (based on block pass-fail)

	Coefficient	Std.Error	Z	p
Intercept	3.1653	0.4683	6.7599	0.0000
SRep Level 2	-0.9938	0.4851	-2.0484	0.0405
SRep Level 3	-2.1150	0.5122	-4.1296	0.0000
BPI (scaled)	-0.7026	0.2103	-3.3405	0.0008
SES (scaled)	0.5842	0.1917	3.0469	0.0023
STM (scaled)	0.5556	0.2020	2.7510	0.0059
Age (scaled)	0.4082	0.2242	1.8207	0.0687
Gender:M	-0.3864	0.3701	-1.0439	0.2965

Table 7

Fixed effects of the optimal Generalized Linear Mixed Model of accurate repetition of target structures in the SRep test by bilingual children. Random effects: Participant, Item

	BPI	SES	STM
Preposition contrast	*		
Verb contrast	.	*	.
Real verb mapping	*	.	*
Novel verb mapping		.	*

Table 8

Predictor significance in the DELV lexical-semantic tests (indicates significance; . indicates a trend)*

	Coefficient	Std.Error	t-value	p
Intercept	0.5249	0.0164	32.0553	0.0000
BPI (scaled)	-0.0349	0.0118	-2.9493	0.0042
Short term memory (scaled)	0.0312	0.0131	2.3740	0.0200
Working memory (scaled)	0.0286	0.0125	2.2903	0.0246
Cognitive flexibility (scaled)	0.0566	0.0127	4.4560	0.0000
Gender: M	-0.0195	0.0235	-0.8272	0.4106

Table 9

Coefficients of the optimal linear regression model fitted to bilingual children composite score in the DELV lexical-semantic tests

	Coefficient	Std.Error	Z	p
Intercept	0.6384	0.0255	24.9904	0.0000
BPI (scaled)	-0.0485	0.0189	-2.5628	0.0122
SES (scaled)	0.0549	0.0200	2.7476	0.0074
Cognitive flexibility (scaled)	0.0717	0.0197	3.6399	0.0005
Gender:M	-0.0170	0.0366	-0.4646	0.6434

Table 10

Coefficients of the optimal Linear Regression Model predicting bilingual children's accuracy score in the discourse semantics task

	Resid. Df	Resid. Dev	Delta	Weight
Cumulative English Exposure	2467	1757.62	0.00	0.37
BPI	2467	1757.85	0.30	0.32
Cumulative English Use	2467	1757.87	1.30	0.19
Current English Exposure	2467	1757.35	2.70	0.09
Current English Use	2468	1758.06	4.70	0.04

Table 11

Model selection for the effect of alternative measures of language experience (listed in column 1) on bilingual children's language proficiency (indexed by the Sentence Repetition score) fit to 2477 observations (i.e., total 2468 degrees of freedom) with 3105 null deviance. The weight indicates the probability that the model is the best one in the set. Delta is the AIC difference of a model compared with the best one.

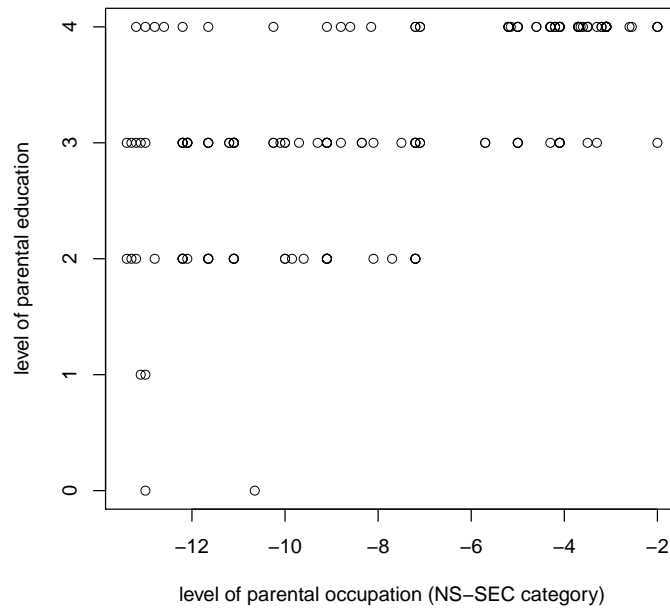


Figure 1. Socio-economic status by education and occupation

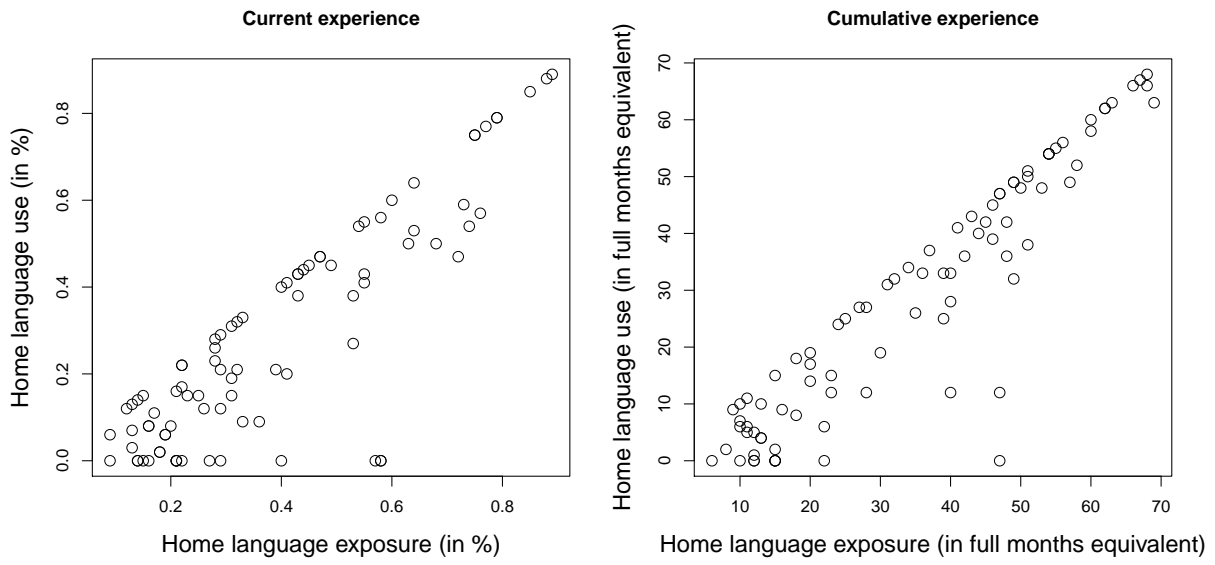


Figure 2. Bilingual children’s current and cumulative experience in their home language (HL)

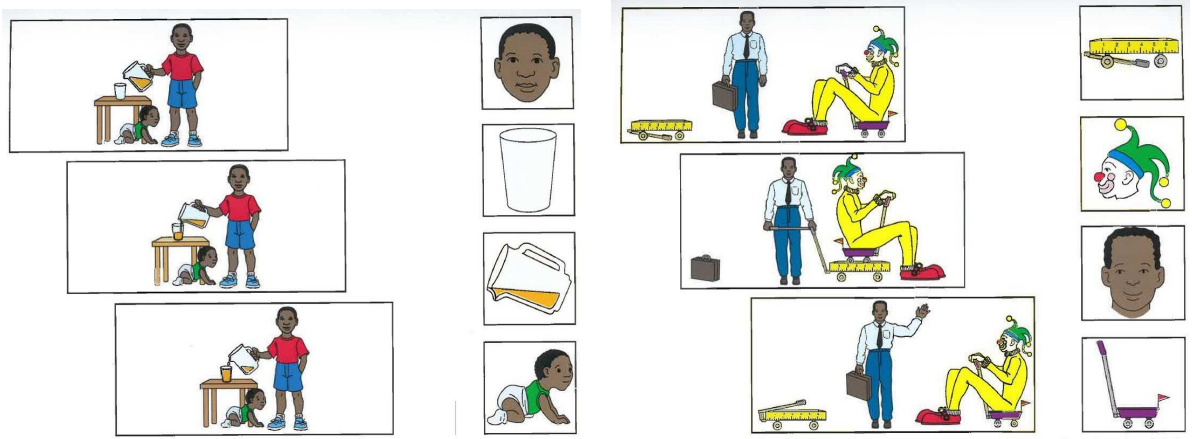


Figure 3. Real verb mapping (left) and novel verb mapping (right)

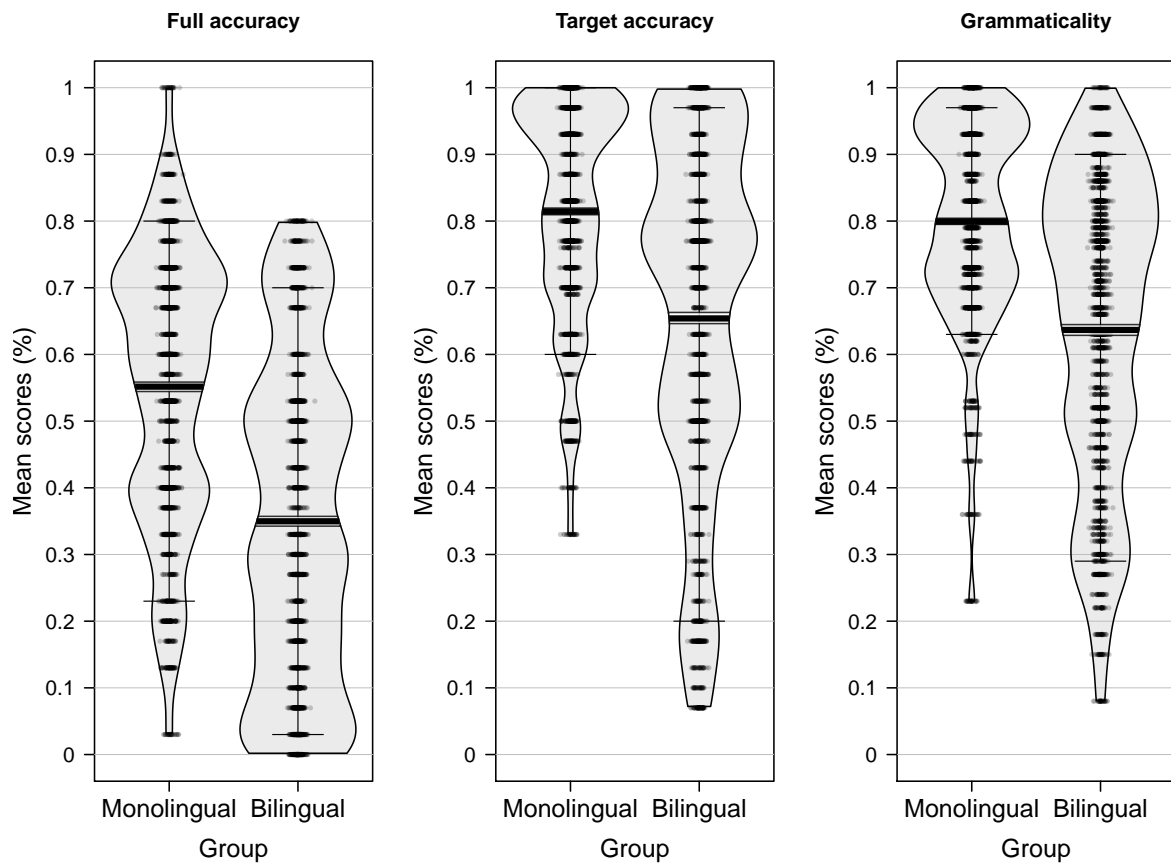


Figure 4. Pirate plots showing the distribution of the raw SRep test scores (according to three different scoring methods). Each plot shows group mean (thick line), confidence intervals (lighter area around the mean) and 10% and 90% quantiles (whiskers), and individual scores.

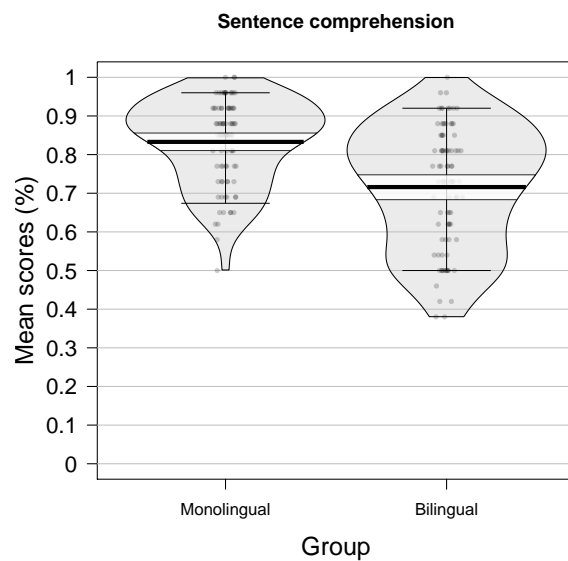


Figure 5. Pirate plot showing individual mean CELF scores, showing group mean (thick line), confidence intervals (lighter area around the mean) and 10% and 90% quantiles (whiskers).

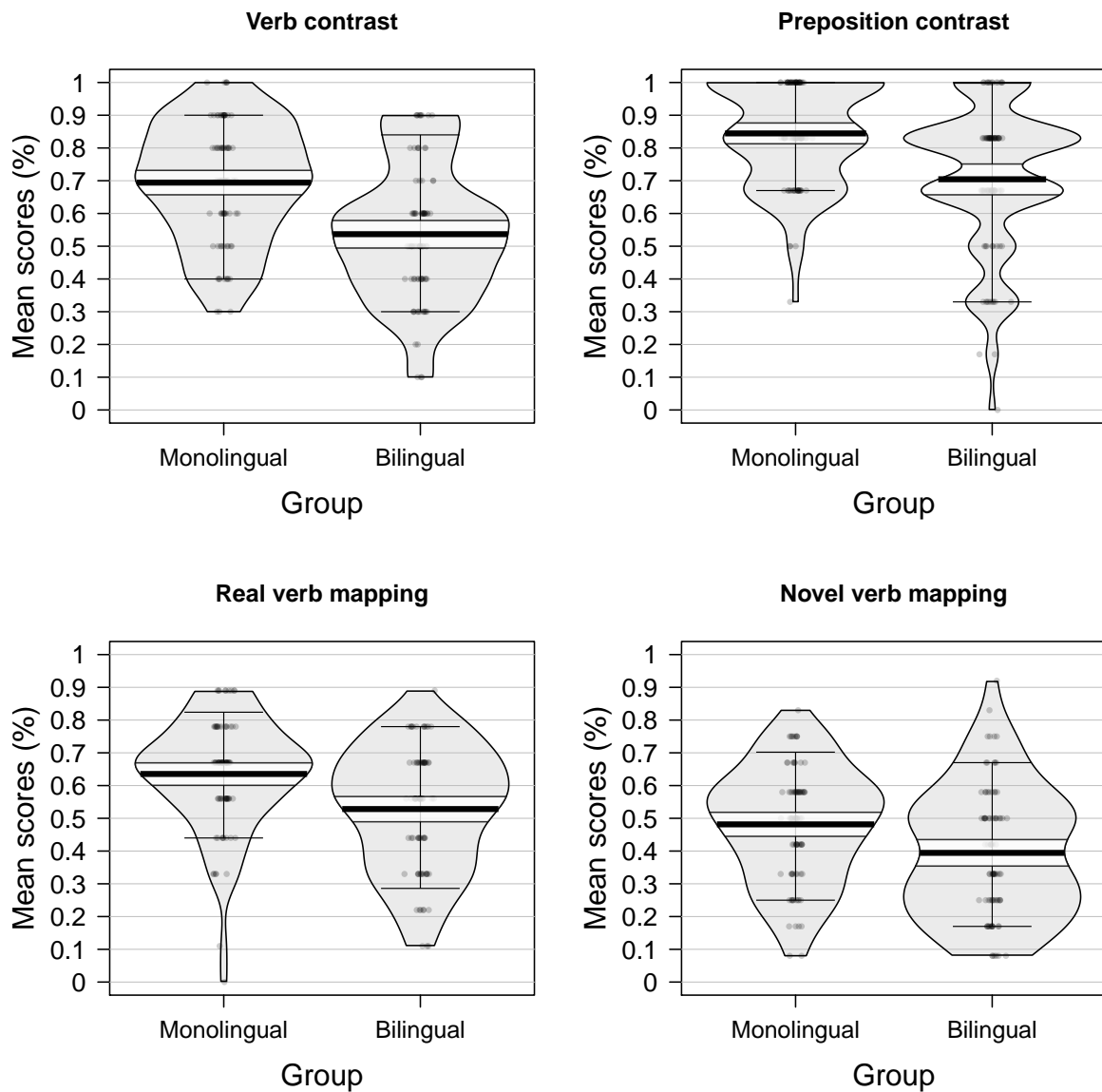


Figure 6. Distribution of the scores on the DELV sub-tests, in bilingual and monolingual children (showing mean, 95% CI and full range)

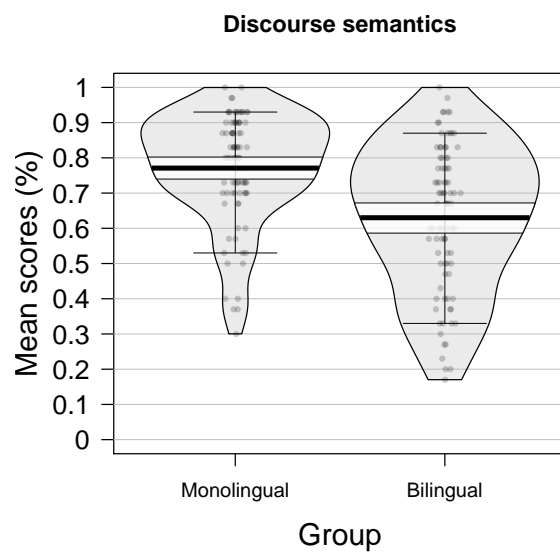


Figure 7. Pirate plots for individual mean scores in the Articles test, showing mean (thick line), confidence intervals (lighter area around the mean) and 10% and 90% quantiles (whiskers).

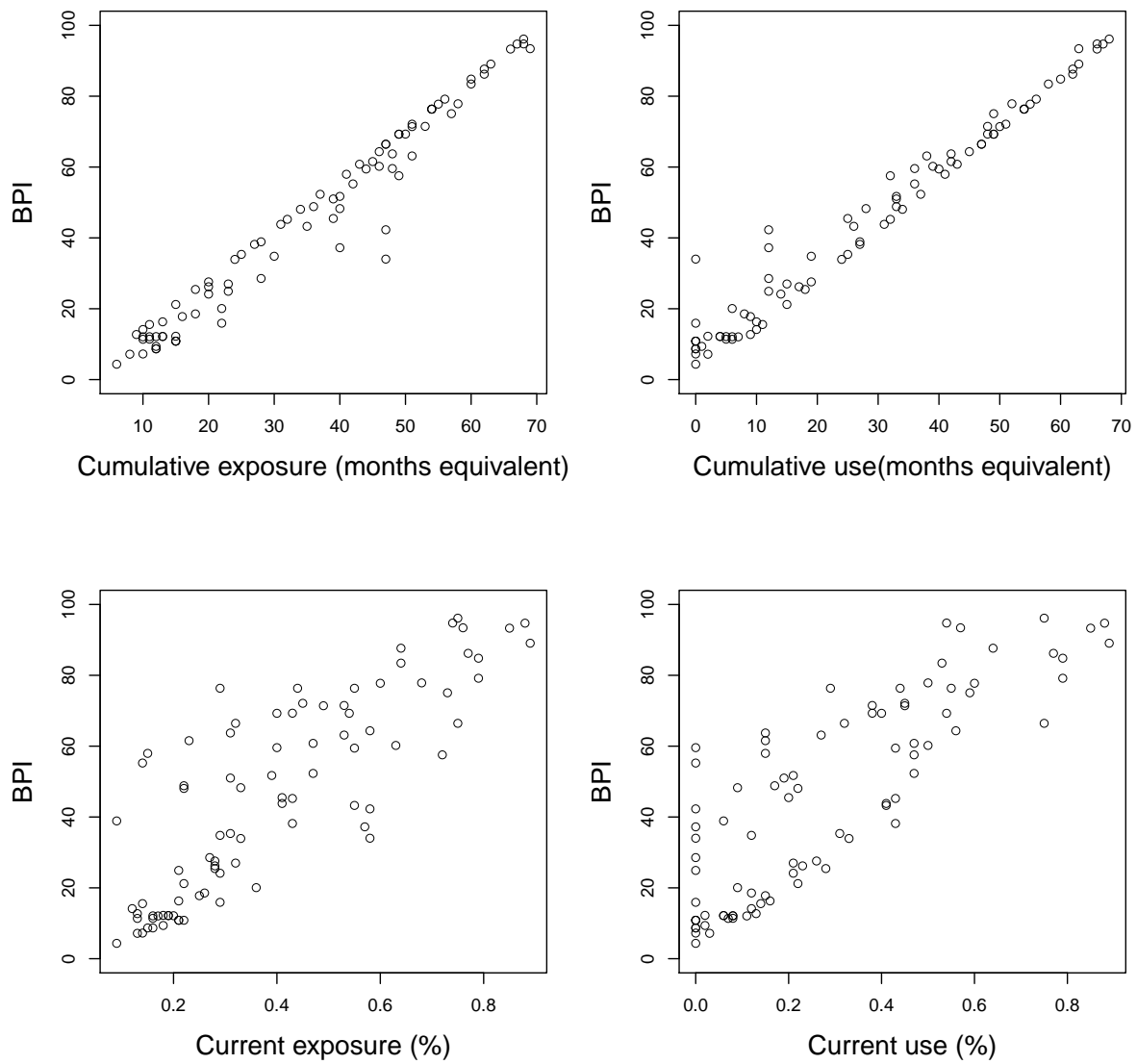


Figure 8. The Bilingualism Profile Index and its relationship with exposure and use measures in the home language. Each dot represents one bilingual child. Monolingual children would score 0 on both axes.

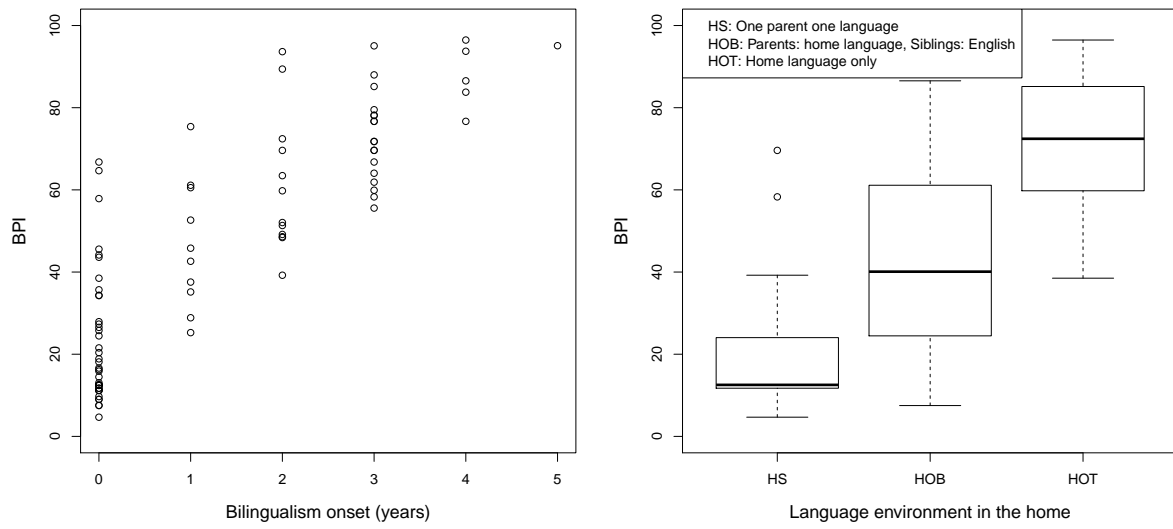


Figure 9. The relationship between the BPI and coarser estimates of bilingual experience: Age of Onset and Language Environment in the Home

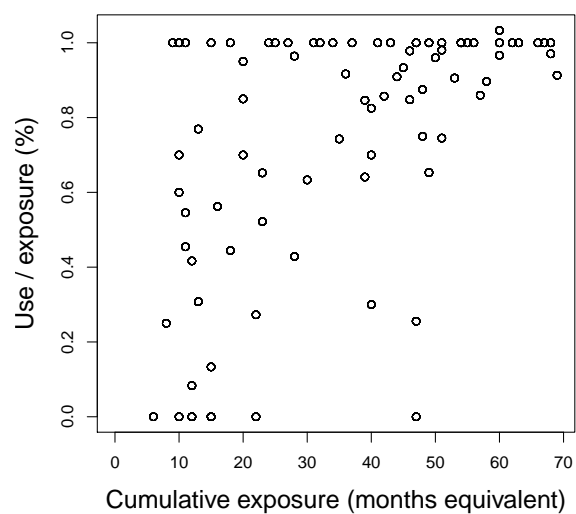


Figure 10. (Im)balance between cumulative use and cumulative exposure of the home language, ranging from an equal amount of both (100% active usage) to no use (total “passivity”)

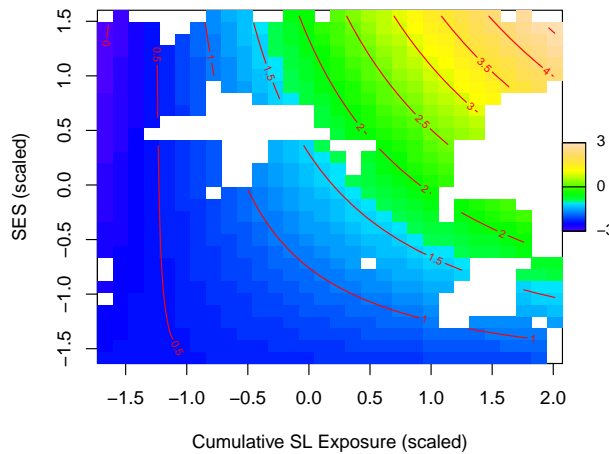


Figure 11. Non-linear interaction between Cumulative SL Exposure and socio-economic status as predictors of bilingual children's accuracy in the sentence repetition (SRep) task estimated through a generalized additive mixed model. Average SRep performance corresponds to 0 on the color gradient.

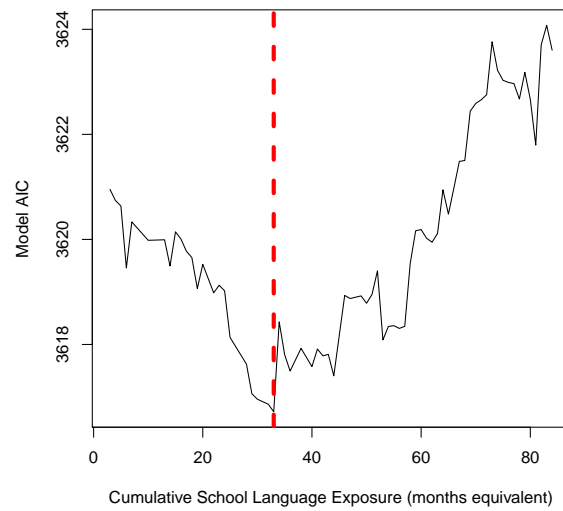


Figure 12. Optimal threshold of cumulative School Language exposure above which bilinguals perform within the monolingual range in the sentence repetition task. The continuous line represents fluctuations in AIC (indicating model fit) depending on where the cut-off point is situated along the SL exposure continuum. The dashed line indicates the lowest AIC, corresponding to the best-fitting model, and hence the optimal cut-off point.

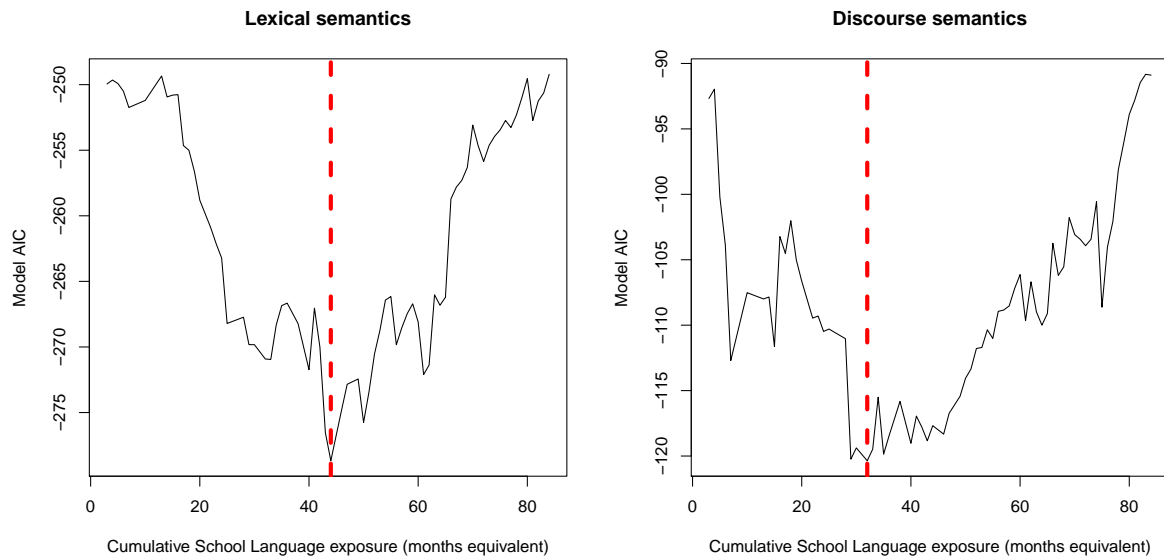


Figure 13. Optimal threshold of cumulative School Language exposure above which bilinguals perform within the monolingual range in the lexical semantics and discourse-semantics tasks. The continuous lines represents fluctuations in AIC (indicating model fit) depending on where the cut-off point is situated along the SL exposure continuum. The dashed lines indicate the lowest AIC, corresponding to the best-fitting models.

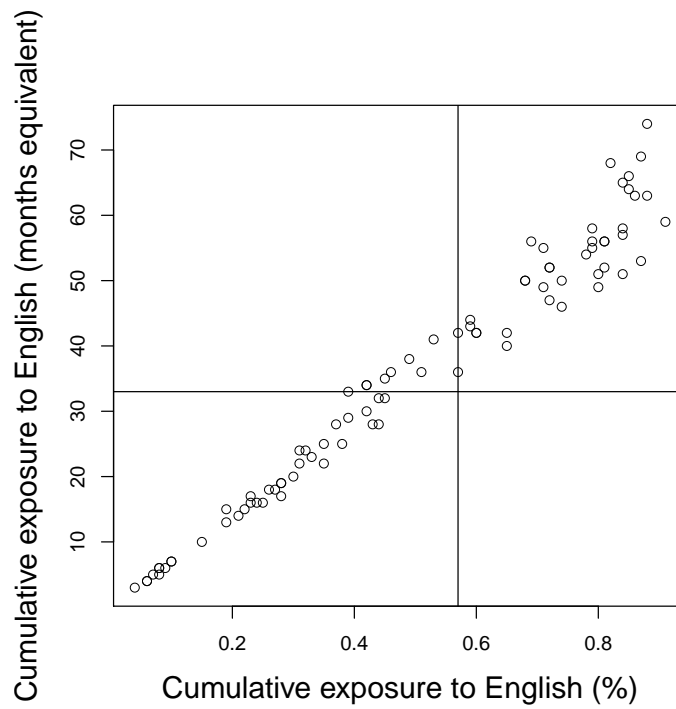


Figure 14. Cumulative exposure to the school language expressed in months equivalent (y-axis) and as a proportion of total language exposure (x-axis). The lines indicate the threshold of exposure required for monolingual-like performance in the SRep test.

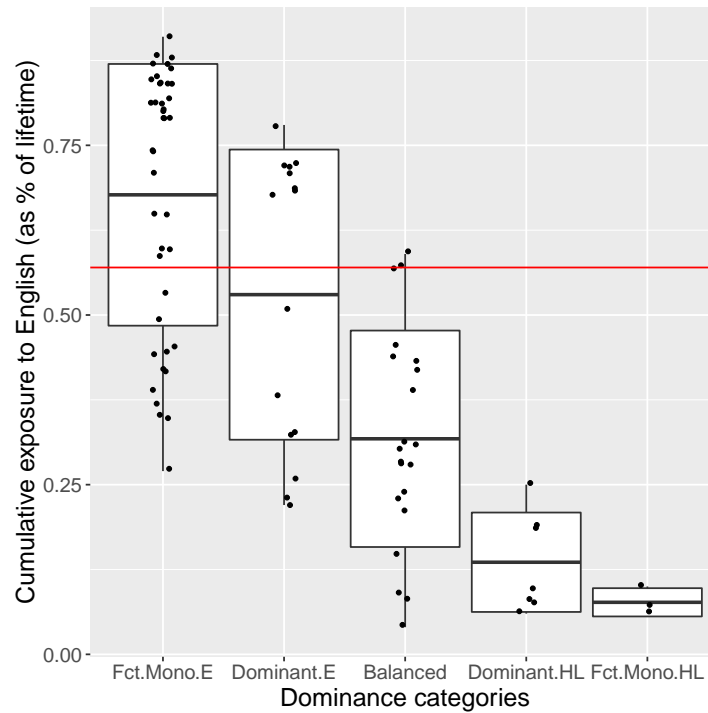


Figure 15. Comparison of Bedore et al's language dominance categories and a continuous measure of language dominance based on cumulative exposure to the school language (in months equivalent). The horizontal line indicates the threshold of exposure required for monolingual-like performance in the SRep test.