- How WEIRD-biased is CHILDES' data on children's linguistic input
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Abstract

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In recent years, the importance of estimating demographic biases in research has become apparent. Here we provide a systematic review of the CHILDES archive, the major source of data on naturalistic recordings of children's linguistic environment. We analyzed the 10 archive at the country and corpus level for four dimensions considered central for language 11 learning: SES, urbanization, family structure and language. We compared these descriptive 12 statistics to world statistics to assess whether the archive was biased in terms of the 13 demographics of the countries represented and the families recorded within them. We found that at the country level, the 47 countries from which there were recordings in CHILDES overrepresented countries with higher educational level; were more urban; and had smaller 16 households with less children. At the corpus level, middle- and higher-class participants were over-represented in relation to the statistics of their own countries. Corpora also 18 included more educated families, with academics being especially over-represented. The 19 corpora were not representative of their countries in terms of urbanization either - with a 20 larger percentage of families residing in urban settings than is overall true for the 21 respective countries. In terms of family structure, nuclear families were more prevalent 22 than in the countries the data was collected in, and - surprisingly - children with no 23 siblings appeared to be under-represented. Last, we found that corpora were linguistically diverse, but we estimate that data in CHILDES under-represents bilingual and multilingual 25 households. We conclude that when generalizing from analysis of data obtained from CHILDES, researchers should acknowledge the potential biases of the archive. 27

28 Keywords: childes, verbal input, infant-directed speech, language, weird

Word count: 8,300 words

How WEIRD-biased is CHILDES' data on children's linguistic input

31 Research highlights

CHILDES should acknowledge these biases.

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We examined CHILDES, the major source of data on children's linguistic
environment for potential sampling bias in terms of SES, urbanization, family structure
and languages. We found that the 47 countries present in CHILDES overrepresented rich,
educated, urbanized countries with small nuclear families. Within these countries, corpora
overrepresented rich, educated, urban and nuclear families - and single-child families, as
well as bilingual families were underrepresented. Interpretation of studies based on

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This paper provides an analysis of the CHILDES database, aiming to assess any
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   potential biases in demographic sampling. We examine the database through the lenses of
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   different social and demographic factors (SES, urbanization, family structure, and
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   language) that influence language learning. Since its foundation in 1984, CHILDES (the
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   Child Language Data Exchange System, MacWhinney, 2000) has been the major source of
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   naturalistic recordings and transcript data for researchers studying language acquisition.
   Naturalistic recordings provide insight into how children acquire language in everyday
   contexts and capture the richness and complexity of language use in everyday
   conversations. They also constitute a valuable and ecologically valid source of data
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   necessary for computational models seeking to reverse engineer language acquisition or
   simulate natural language development. Additionally, they form the foundation for many
   influential concepts in developmental science. For instance, they helped establish
   correlations between caregiver behavior, home environment, and child development, such as
   the link between socio-economic status and language and literacy (Hoff, 2013). Last, they
   inspire theories that help understand the connection between language input quantity and
   language development (Christiansen et al., 2022). However, naturalistic recordings may
   also contain potential biases. One notable concern is that participants might not represent
   the diverse socio-economic backgrounds, cultures, or linguistic environments children
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   typically experience. As a result, the generalizability of such research may be limited,
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   necessitating cautious interpretation. Researchers in many fields are increasingly aware of
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   the bias towards WEIRD populations (Western, Educated, Industrial, Rich, Democratic,
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   Henrich 2010) in the samples they study (Nielsen et al., 2017; Moriguchi, 2021; see Cychosz
   & Cristia, 2022, Singh et al., 2023). Recent calls have been made to diversify research in
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   psychology, cognitive, and developmental science to address this issue (Majid & Levinson,
   2010; Kidd & Garcia, 2022; Blasi et al. 2022). For instance, Kidd and Garcia (2022)
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   systematically sampled publications from child-language journals. They revealed biases
   towards specific continents (North America and Europe) and languages (English. Spanish,
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French), highlighting the need for increased diversity in populations and languages studied.

Similarly, Blasi et al. (2022) emphasize the potential consequences of generalizing

observations derived solely from English speakers and how fit they are to represent our

entire human species. [TRANSITION] Every area of research conceptualizes different

dimensions relevant to explaining variance for a given research topic. In what follows we

present the four most relevant dimensions thought to play a role in the development of

child language, particularly related to naturalistic recordings: Socioeconomic Status,

Urbanization, Family Structure, and Language.

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Socioeconomic Status. Decades of research have examined the linguistic 74 differences among families with varying socio-economic status (SES). In the developmental 75 literature, this is primarily indexed by parents' education (Ensminger, Fothergill, 76 Bornstein, & Bradley, 2003; Hoff, 2003a), but can also be indexed otherwise, such as by parental income, occupation, or a composite measure of those three (e.g. Hollingshead, 78 1975). It is beyond the scope of this paper to detail all the theories that attempt to account for the complex causal pathways that may connect SES to children's language environments. We recommend Rowe (2018) as a starting point for readers interested in this 81 literature, along with Golinkoff et al. (2019) and Sperry et al. (2019) for diverse theoretical perspectives. It should be noted, however, that some of this literature has been found to reflect majority biases, including what kinds of language input are counted, and what features of linguistic experiences are valued (Sperry, et al., 2019; Scaff et al., 2023). Without desiring to take a stance on how SES and language environments relate to each other, we merely indicate here that SES is undoubtedly one of the factors that has been repeatedly studied in the context of early language acquisition, including in terms of input description. For example, Hoff (2003a) compared the speech of high- versus low-SES American mothers. College-educated parents produced more utterances to their child, with more diverse vocabulary, and longer phrases, and higher number of utterances continuing a topic the child had brought up. Similar findings can be seen in other studies (Hart &

Risley, 1995; Hoff-Ginsberg 1990; Hoff, 2003b; Huttenlocher, et al., 2002, 2007; see also Dailey & Bergelson, 2022; Piot et al., 2021; for meta-analyses supporting the link; and Bergelson et al., 2023 for a large-scale study finding non-significant SES effects).

Urbanization is the process of moving from rural to urban areas Urbanization. 96 along with noticeable changes in job opportunities and living conditions. It involves the 97 growth and development of cities, leading to increased access to infrastructure and 98 amenities. Studies indicate that children exposed to higher levels of noise pollution in urban areas exhibit a reduction in cortical thickness in the left IFG, an area of the brain 100 associated with language development (Simon et al., 2022). Within the general theoretical 101 framework of language socialization, there have been proposals suggesting that societies 102 varying in their urbanization process have differing views and values about the role of 103 children in conversations, and more generally in the community (e.g., Sharma & Levine, 104 1998; Richman et al., 1992; Draper & Harpending, 1987; Keller, 2012). For instance, Keller 105 (2012) discusses three prototypical cases: urban, rural, and hybrid. These three groups 106 differ in terms of their goals for children, with urban families aiming for child psychological 107 independence, rural families for child physical autonomy and interdependence, and hybrid 108 families aiming for some mix across these values. Vogt et al. (2020) employ this conceptual classification to interpret their results on multimodal language use across three samples, 110 namely urban Dutch, urban Mozambique, and rural Mozambique, finding that the number 111 of gestures, gesture-speech alignment, and gesture types all vary across the three groups in 112 ways that can be related to Keller's typology. 113

Family structure. Family structure refers to the arrangement within a household, forming the basis of a family unit. This dimension encompasses aspects such as the number of siblings, birth order, the number of caregivers in the household, and the number of individuals sharing or competing for household resources (including caregiving attention); each of which has a significant impact on child and language development (e.g., Blake, 1981; Duncan & Paradis, 2020; Havron et al., 2019, 2022, Hoff, 1993; Tomasello et al.,

1995; Bornstein et al., 2019). For example, birth order effects reveal that children with 120 older siblings show lower language skills than first-born children in various cultures (e.g., 121 Pyere et al., 2016 in France; Havron et al., 2022 for Singapore; Zambrana, et al., 2012 for 122 Norway). [Another example], in middle-class Euro-American families, parents typically 123 assume primary responsibility for children, often focusing on the mother as the primary 124 caregiver (e.g. Bakermans-Kranenburg et al., 2004; Huttenlocher et al., 2010; Ispa et al., 125 2004; Pan et al., 2005). However, certain cultures, like Turkish families described by 126 Isleyen (2021), may adopt a different approach, with nuclear families living in separate 127 apartments but sharing common spaces and caregiving responsibilities, resulting in 128 extensive support networks. [Importance of family size. BUT Family size does not seem to 129 affect language acquisition when controlling for factors such as SES and birth order (see 130 Blake, 1981). ### Languages Characterizing the diversity of participant samples in terms 131 of language (Blasi et al, 2022; Kidd & Garcia, 2022) is an important factor in language 132 acquisition, as variations in language exposure and language use among different groups 133 allow to explore how purely linguistic factors shape and influence the development of 134 language skills in children. Another major dimension and entire sub-field in developmental 135 science is the study of bilingualism or multilingualism (for an introductory review see XX, 136 also McCabe et al., 2013). We refer to lingual status as the variable that distinguishes 137 monolingualism from bilingualism and other cases. 138

The current study. Here, we provide a systematic analysis of the naturalistic
speech corpora of the CHILDES database by quantifying the diversity of each dimension
presented above (see Tables 1 and 2). Though some of these dimensions overlap with each
other (See Figures SM.X and SM.x in the supplementary materials for illustrations), we
decided to illustrate each, as best as possible, independently. Although valuable insights
can be gleaned from corpora on clinical populations and elicitation tasks, we deliberately
excluded them from our review. This is because they introduce additional biases. For
example, the inclusion of clinical populations may highlight differences in the prevalence

and access to diagnosis of language disorders in different regions or countries. Additionally, 147 the structure of elicitation tasks may not accurately reflect spontaneous language use in 148 everyday conversations. Our systematic analysis follows three steps. First, we screened the 149 CHILDES database, excluding the clinical and task-driven recordings. Second, we 150 extracted information related to the four central factors in language acquisition described 151 in the Introduction. Finally, we follow Ghai's (2022) recommendation to improve the 152 description of diversity in behavioral sciences by looking at different levels: from a 153 macro-level, with broad country comparisons to a corpus-level, where we delved into 154 individual corpora to gain a more detailed and nuanced understanding of the data. 155

Methods 156

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Analyses and visualizations were carried out using R (version 4.1.2, R Core Team, 2020) and ggplot2 (Wickham, 2016). Data, scripts, and online Supplementary Materials are 158 available on OSF [https://osf.io/q9w82/?view only=a013f1b25b8c4556b8248f12870402c9].

Inclusion criteria. We excluded the following corpora: a) clinical populations or 160 non-typically developing children, b) structured tasks such as toy narratives, personal 161 narratives, frog stories, picture or movie descriptions, structured storytelling, and other 162 elicitation tasks; c) only child or adult speech without a conversational partner; and d) 163 non-naturalistic setups, such as recordings conducted exclusively in a lab environment or in 164 a diary format. ### Screening Following a thorough examination of each corpus, which 165 involved reviewing the corpus description available on the CHILDES website, checking for 166 any accompanying references such as articles, book chapters, or dissertations, and 167 conducting spot-checks on associated transcripts, we identified 180 corpora that met our 168 inclusion criteria mentioned above. For a detailed breakdown of the included corpora, 169 please refer to the flowchart in Supplementary Materials. 170

Descriptive-statistics. Descriptive statistics Firstly, we present an overview of the 171 countries within the CHILDES database, specifically the distribution of participants by 172

continent (Table 2) and geographical location (Figure 1) and examining three out of our 173 four key dimensions: SES, urbanization, family structure (Figure 2). Our goal is to provide 174 insights into the representativeness of our CHILDES sub-sample when compared to global 175 statistics. Figure 2 draws data from official sources such as the World Bank, Our World in 176 Data (WDI), and the United Nations (UN). It is noteworthy that information on the 177 Language dimension is omitted due to the lack of official world statistics on 178 multilingualism. This limitation arises from the exclusion of certain languages or dialects 179 in official country counts (REF). For additional details on Figure 2 and the corresponding 180 variables, please refer to the Supplementary Materials (SMX). 181

Secondly, we offer corpus-level statistics to assess the representativeness of our 182 sub-sample of CHILDES in a more detailed manner across our four dimensions: SES, 183 urbanization, family structure, and languages (See Table 1 for the complete list of variables 184 and definitions). Data was extracted from the provided sources mentioned in CHILDES 185 such as articles, book chapters, dissertations, and transcripts to pre-fill the categories. 186 Corpus curators were contacted to request missing or incomplete information, and an 187 online table was provided to facilitate data entry. Over a third of the contacted curators 188 (XX corpora, XX%) provided additional data or confirmed missing information. 189 Unfortunately, curators for 180 corpora (XX%) could not be contacted due to unresponsive 190 email addresses (38), or the curator's passing (12). 191

Table 1

Dimension	Macro.level.variable	
SES	Percent of the population completing lower secondary school*	
SES	GDP per capita (log 10)	
Urbanization	Percent of the population living in urban areas	WDI from the World
Family structure	Average household size	

Table 1 continued

Dimension	Macro.level.variable
Family structure	Average number of member under 15 in households
Language	NA

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Table 2

Definition of the corpus-level variables

Dimension	Corpus.level.variable	
SES	Parents' socio-economic status	
SES	Parents' education level	Highest le
SES	Parents' occupation	
Urb	Type of community	
Family structure	Household composition	Whether the family was composed primarily of care
Family structure	Percent children with sibling(s)	
Family structure	Average number of siblings	How
Language	Language(s) spoken	
Language	Lingual status	Whether more than one language is

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194 Results

The 180 corpora included in this study represent 48 different countries across all populated continents. Out of the 48 different countries or areas included in the sample, 28 of them belong to the OECD (Organisation for Economic Co-operation and Development).

This corresponds to 149 out of 180 corpora or to 82.78% of this sub-sample of CHILDES.

OECD countries represent 19.5% of world countries.

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Table 3 Descriptives

Continent	N.corpora	N.total.children	N.children.growing.in.continent
Africa	3.00	16.00	X
Asia	32.00	602.00	X
North America	34.00	177.00	X
Latin America	5.00	6.00	X
Europe	105.00	1,091.00	X
Oceania	1.00	5.00	X

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We assessed the extent to which the countries with data in our sub-sample of CHILDES were a representative sample of countries in the world. Density plots are portrayed in Figure 2.

By comparing our sub-sample of CHILDES to the world statistics using unpaired samples t-tests without assuming equality of variance (Welch's t). Countries in our sub-sample of CHILDES had a higher proportion of the population completing lower secondary school than the world wide sample (% compl. LSS, t(130.41)=-6.19, p = 0); they were more urban (% urban, t(79.48)=-3.44, p = 0); richer (log GDP per capita, t(102.35)=-6.02, p = 0) and had smaller households (average household size, t(118.80)=7.12, p = 0) and less average number of members under the age of 15 (average under 15 size, t(26.12)=5.01, p = 0).

Participants by Country

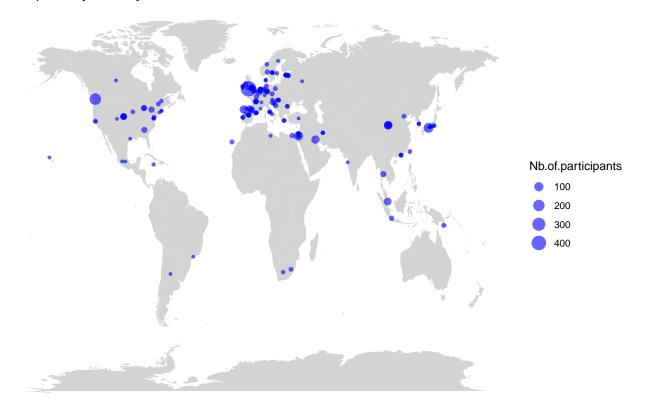


Figure 1. World map showing corpus. Size proportional to number of children per corpus

213 Corpus-level

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We then investigated at the corpus level the variables described in Table 2. The information collected allows us to have more detail about the characteristics of the individual families that compose our sub-sample of CHILDES. However, it is important to acknowledge that for the majority of the corpora, this information was missing, not known, or not provided.

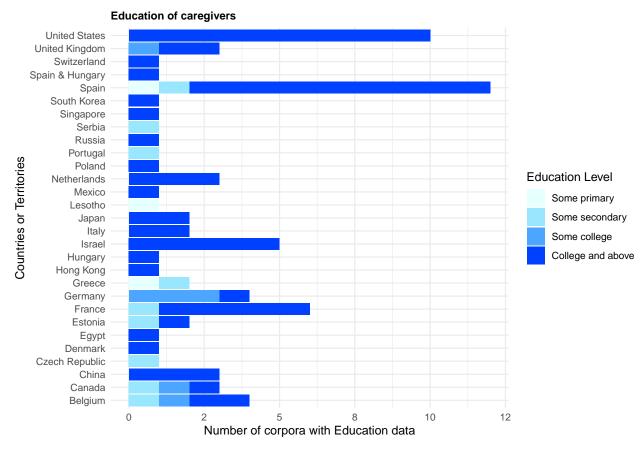
SES. Of the 102 (43%) corpora, 5 were described as having low SES; 16 were described as spanning both lower and middle or higher SES; and 81 were described as middle or higher SES exclusively (79%). Given that most countries represented in our sub-sample of CHILDES are in the OECD, we can compare this proportion with the proportion of the population in these countries that are middle class. According to a 2016

report, "Almost two-thirds of people live in middle-income households in OECD countries", for whom "household net income [is] between 0.75 and 2 times the median". Thus, middle and higher-class participants appear to be over-represented in our sub-sample of CHILDES data.

For socioeconomic status, there were 106 missing values (33%). Of the remaining 215 228 samples, were described as having low SES; were described as spanning both lower and 229 middle or higher SES; and were described as middle or higher SES exclusively. Given that 230 most countries represented in CHILDES are in the Organization for Economic Cooperation 231 and Development (OECD, 0 out of the 321 corpora), we can compare this proportion with 232 the proportion of the population in these countries that are middle class. According to a 233 2016 report, "Almost two-thirds of people live in middle-income households in OECD 234 countries", for whom "household net income [is] between 0.75 and 2 times the median". 235 Thus, middle and higher class participants appear to be over-represented in CHILDES 236 data, composing 81% of available data. 237

76% of the corpora (n = 58) include children whose parents had at 238 least a graduate, if not a postgraduate, degree. 4% (n=3) had at least some parents with 230 primary-level education: 12% (n=9) had parents with secondary school education as the 240 lower bound of the education range, and a further 8% (n=6) had some college as the lower 241 bound. 4 corpora were described as "diverse", without clarifying the range of education 242 covered. These numbers do not accurately represent the demographics of the countries 243 they were obtained from. For instance, while the U.S. Census Bureau reported that only 244 36% of the adult population held a bachelor's degree or higher in 2020, our data indicates that 100% of the corpora of the parents from the United States had a college education or higher. As seen in Figure 3, the same result is seen for corpus from China, Denmark, Egypt, Hong Kong, Hungary, Israel, Italy, Japan, Mexico, The Netherlands, Poland, Russia, Singapore, South Korea, Switzerland and the United States where 100% of parents 249 with data are college-educated or above. Thus, it seems that our sub-samples of CHILDES 250

are very skewed toward higher-educated parents.



Occupation Professions were overall varied. The majority, comprising62% (n=52),
was associated with the field of education. Notably, within this category, 56% (n=47) of
the individuals were identified as parents with professions linked to graduate-level
education. This included roles such as Master's or Ph.D. students, professors, linguists,
researchers, scientists, and academics. Some of the other occupations reported were
psychologists, speech therapist or home makers.

Urbanization. 78% of the corpora (n=51) corpora were described as industrialized or urban, and an additional one as both rural and urban. Only 7 corpora were described as farming or rural. In these same countries, the proportion of the population residing in urban settings was 76%, suggesting that samples were not representative of their countries in terms of rural versus urban settings either.

Family structure.

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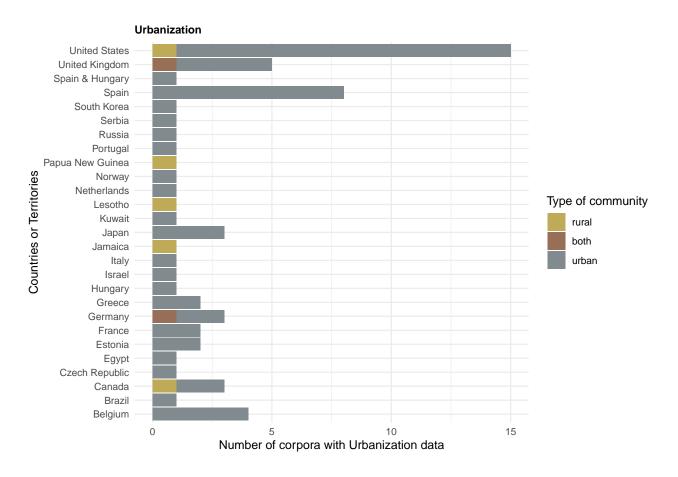


Figure 2. Urbanization by country.

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Nuclear households. As for household composition, 60 corpora (87%) were based on nuclear families; 7 extended families(10%); and in 2 corpora the structure was varied (3%). We compare side by side the average per country of the nuclear households (the sum of the percentages of couples with children households, and single parents with children households; United Nations, 2022).

Sibling presence. A majority of corpora in our sub-sample of CHILDES include children who have siblings. In fact, only 28% of corpora (among the 93 corpora having information on siblings) were constituted exclusively of children with no siblings, and the remaining had at least one sibling, with the overall average being 0.8 siblings. Since 83% of countries in CHILDES are in the OECD, we draw a comparison point for such countries: 46% of children had no siblings in OECD countries according to 2015 data. In this sense,

children with no siblings appear to be under-represented in CHILDES.

Languages. We had two variables of interest here: language spoken in the corpus and lingual status. A total of 62 different languages or language combinations (for bilingual and multilingual children) were reportedly spoken in the corpora.

Table 4

Total number of participants per language,

Monolingual_Language	Total_Participants	${\bf Multilingual_Languages}$	Total_	_Partio
Afrikaans	2.00	Dutch/English, Dutch/French		
Arabic (Egyptian or Kuwaiti)	80.00	Dutch/Italian		
Basque	46.00	Spanish/Catalan		
Cantonese	8.00	Spanish/Galician		
Catalan	17.00	English/Cantonese		
Cree	1.00	English/French		
Croatian	3.00	English/Hebrew		
Czech	6.00	English/Japanese		
Danish	2.00	English/Japanese/Danish		
Dutch	23.00	English/Mandarin		
English	210.00	English/Mandarin/Cantonese		
Estonian	32.00	English/Russian		
Farsi	5.00	English/Spanish		
French	46.00	French/Russian		
German	46.00	Hungarian/Catalan/Spanish		
Greek	6.00	Hungarian/Farsi/English		
Hebrew	122.00	German/Spanish		
Hungarian	8.00	Italian/Japanese		

Table 4 continued

Monolingual_Language	Total_Participants	${\bf Multilingual_Languages}$	Total_Partic
Icelandic	1.00	Italian/German	
Indonesian	8.00	Portuguese/Swedish/English	
Irish	7.00	NA	
Italian	8.00	NA	
Jamaican	2.00	NA	
Japanese	148.00	NA	
Korean	4.00	NA	
Mandarin	157.00	NA	
Norwegian	11.00	NA	
Nungon	5.00	NA	
Polish	4.00	NA	
Portuguese (Brazilian or European)	9.00	NA	
Romanian	6.00	NA	
Russian	2.00	NA	
Serbian	8.00	NA	
Sesotho	4.00	NA	
Slovenian	20.00	NA	
Spanish	75.00	NA	
Swedish	9.00	NA	
Taiwanese	4.00	NA	
Tamil	1.00	NA	
Thai	18.00	NA	
Turkish	1.00	NA	
Welsh	475.00	NA	

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About a third (32%) of the included corpora that had available data for this variable (N = 110) were not monolingual. It is hard to find reliable estimates of the percentage of the population which is not monolingual in the world or in the countries represented in our sub-sample of CHILDES, but for instance, in Europe in 2016, 65% of adults reported knowing multiple languages (Eurostat, 2022). According to such estimates, even if samples are linguistically diverse, it would appear that input data in our sub-sample of CHILDES under-represents bilingual and multilingual households.

288 Discussion

References