# Web-CDI: A system for online administration of the MacArthur-Bates 1 Communicative Development Inventories

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10 Abstract

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Word count: X

Understanding the mechanisms which drive variation in children's language acquisition 11 requires large, population-representative datasets of children's word learning across 12 development. Parent report measures such as the MacArthur-Bates Communicative 13 Development Inventories (CDI) are commonly used to accrue such data, but the traditional paper-based forms make the curation of large datasets logistically challenging, and tend to 15 rely on convenience samples located close in proximity to major research institutions. Here, 16 we introduce Web-CDI, a web-based tool which allows researchers to collect CDI data 17 online. Web-CDI contains functionality to collect and manage longitudinal data, share links, and download standardized vocabulary scores. To date, over 3,500 valid Web-CDI administrations have been completed. General trends found in past norming studies of the CDI are present in data collected from Web-CDI: scores of children's productive vocabulary grow with age, female children show a slightly faster rate of vocabulary growth in early childhood, and participants with higher levels of education attainment report 23 slightly higher vocabulary production scores. We also report results from an effort to 24 oversample non-white, lower-SES participants (N = 241), which showed similar 25 demographic trends to the full sample but which had a high exclusion rate. Implications 26 and challenges for the collection of large, population-representative datasets using 27 Web-CDI in future research are discussed.

Keywords: vocabulary development, parent report, socioeconomic status

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Children vary tremendously in their vocabulary development (Frank, Braginsky,
Yurovsky, & Marchman, 2021). Characterizing this variability is central to understanding
the mechanisms that drive early language acquisition, yet capturing this variation in broad,
diverse samples of children has been a significant challenge for cognitive scientists for
decades. The MacArthur-Bates Communicative Development Inventory (MB-CDI, or CDI
for short) is a commonly-used parent report instrument for assessing vocabulary
development in early childhood (Fenson et al., 2007) that was introduced in part to create
a cost-effective method for measuring variability across individuals.

In this paper, we introduce a web-based tool, Web-CDI, developed to address the
need for collecting CDI data in an online format. Web-CDI allows researchers to increase
the convenience of CDI administration, further decrease costs associated with data
collection and entry, and access participant samples that have traditionally been difficult to
reach in language development research. Our purpose in this paper is twofold: first, we
describe Web-CDI as a platform which streamlines the process of collecting MB-CDI data
and collates the data in a way that facilitates the creation of large-scale, multisite
collaborative datasets. Second, we profile usage of Web-CDI thus far, with a particular
focus on broadening the reach of traditional paper-based methods of collecting vocabulary
development data.

We begin by discussing parent report as a powerful method by which to address the challenge of measuring early language outcomes, as well as previous online parent report instruments. We then describe Web-CDI and its use from the perspective of participants and researchers. Finally, we report on our use of Web-CDI thus far and discuss the potential of Web-CDI to acquire vocabulary data from diverse, population-representative samples. We end with some challenges for future research.

### The Importance of Parent Report Data

Gaining empirical traction on variation in children's early language requires reliable 58 and valid methods for measuring language abilities, especially in early childhood (8 to 30 59 months). Parent report is a mainstay in this domain. Parent reports are based on their daily experiences with the child, which are much more extensive than a researcher or clinician can generally obtain. Moreover, they are less likely to be influenced by factors that may mask a child's true ability in the laboratory or clinic (e.g., shyness; Frank et al. (2021)), and parents are remarkably accurate at reporting their children's language, especially in the first two years of life (CITE). One widely used set of parent-report instruments is the MacArthur-Bates Communicative Development Inventories, originally designed for children learning American English (Fenson et al., 2007). The American English CDIs come in two versions, Words & Gestures for children 8 to 18 months, focusing on word comprehension and production, as well as gesture use, and Words & Sentences, for children 16 to 30 months, focusing on word production and sentence 70 structure. Together, these instruments allow for a comprehensive picture of milestones that 71 characterize language development in the first 2½ years of life.

A substantial body of evidence suggests that these instruments are both reliable and valid (e.g., Fenson et al., 1994, 2007) leading to their widespread use in thousands of research studies over the last few decades. Indeed, the popularity of the American English and Spanish CDI instruments has meant that many teams around the world have adapted the CDI format to the particular language and community (Dale, 2015). Importantly, these adaptations are not simply translations of the original form but rather incorporate the specific features of different languages and cultures, since linguistic variability exists even among cultures that share a native language (e.g., Cheerios are more common in American than British homes, so age of acquisition of this word may differ substantially). To date there are now more than 100 adaptations for languages around the globe.

Initial large-scale work to establish the normative datasets for the American English
CDI not only provided key benchmarks for determining children's progress, but also
documented the extensive individual differences that characterize early language learning
during this critical period of development (Bates et al., 1994; Fenson et al., 1994).
Understanding the origins and consequences of this variability remains an important
empirical and theoretical endeavor that has informed critical insights in the field (e.g.,
Bates & Goodman, 2001; Bornstein & Putnick, 2012; see also, Frank et al., 2021). The
popularity of the instruments has remained strong over the years, leading to the
development of extensions of the methodology to alternative formats, e.g., short forms
(Fenson et al., 2000).

While the reliability and validity of these instruments is well-established (Fenson et al., 2007) for the American English versions of the forms, existing norming samples are skewed toward families with more years of formal education and away from non-White groups. Representation in the norming samples is generally restricted to families living on the US east and west coasts (CITE FOR THIS). Further, although paper survey administration is a time-tested method, increasingly researchers and participants would prefer to use an electronic method to administer and fill CDI forms, obviating the need to track (and sometimes mail) paper forms, and the need to key in hundreds of item-wise responses for each child.

Here, we report on our recent efforts to create and distribute a web-based version of
the MacArthur-Bates CDIs in order to address some of the limitations of the standard
paper versions. Online administration of the CDI is not a novel innovation – a variety of
research groups have created purpose-build platforms for administering the CDI in
particular languages. For example, Kristoffersen et al. (2013) collected a large normative
sample of Norwegian CDIs using a custom online platform. Similarly, the Slovak adaptation
of the CDI uses an online administration format (CHECK CITE HERE). And many
groups have used general purpose survey software such as Qualtrics and Survey Monkey to

administer CDIs and variants online (e.g., Caselli, Lieberman, and Pyers (2020)). The 110 innovation of Web-CDI is to provide a comprehensive researcher management interface for 111 the administration of a wide range of CDI forms, allowing researchers to manage 112 longitudinal administrations, download standardized scores, and share links easily, all while 113 satisfying strong guarantees regarding privacy and anonymity. Moreover, a key benefit of a 114 unified data collection and storage system such as Web-CDI is that data from disparate 115 sources are combined into a single repository, far reducing the overhead efforts associated 116 with bringing together data collected using paper forms by researchers across the world. 117

# 18 Introducing Web-CDI

Web-CDI is our web-based platform for CDI administration and management. 119 Parents are recruited by either receiving an individual URL directly from a researcher or 120 by interacting with a targeted social media advertisement using general-purpose URLs. 121 Web-CDI serves as a low-risk method that allows researchers to communicate with families 122 electronically, facilitating access to families in areas distant from an academic institution 123 and eliminating costly mailings or laboratory visits. It also permits other institutions to 124 use Web-CDI as a resource, while still allowing members of the CDI Advisory Board to 125 access and analyze the resulting data with the researchers' permissions. Since 2018, more 126 than 3000 CDIs have been collected via 15 research groups throughout the US, 127 demonstrating the potential for large-scale data collection and aggregation.

Below, we outline how Web-CDI is used. We begin by detailing the consent
obtention process and participant experience. Second, we describe the interface that
researchers use to collect data using Web-CDI, specifying a number of common use cases
for the platform. Lastly, we briefly discuss the administrator role.

## $Participant\ interface$

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Participants can complete the Web-CDI on a variety of devices, including personal 134 computers and tablets: Web-CDI can be administered on a smartphone, although the 135 experience is not as ideal for the user due to the length of the survey. When a participant 136 clicks a link to the form, they are directed to a website displaying their own personal 137 administration of the Web-CDI, regardless of whether the link was participant-specific or 138 general-purpose. In some cases, they may be asked to read and accept a waiver of consent 139 documentation, depending on whether the researcher has chosen to use that feature (see 140 also Researcher Interface below). 141

Demographics. The parent is then asked to provide demographic information about 142 their family and any health conditions that might impact their child's vocabulary 143 development. The specific demographic questions asked of participants can be adjusted to 144 vary between different versions of the form, allowing researchers to tailor the demographic questions to local norms<sup>1</sup>. Researchers can customize the presentation of these demographic questions in three ways. First, they can elect to show all of the demographics items on the landing page or they elect to present the majority of these questions at the end of the instrument. This choice is provided because some pilot work in the United Kingdom indicated that answering questions regarding personal health information at the beginning 150 may deter participants from completing the instrument. Second, certain demographic 151 questions can be asked at both the beginning and the end of the form to serve as validity 152 checks, providing a check that can be used to screen for hasty or illegitimate completions. 153 Third, researchers can tailor the questions to the societal and cultural context of their 154 participants (e.g. country-specific education level descriptors and income categories). 155

Instructions. After completing the first demographics page, participants are

<sup>&</sup>lt;sup>1</sup> For example, the Dutch CDI omits questions about participant ethnicity, since census data in the Netherlands does not include ethnicity.

directed to the instructions that are appropriate for either the Words & Gestures or Words 157 & Sentences version (see Figure 1). At the top of the page are general instructions that 158 inform participants that they should expect the study to take at least 30 minutes and that 159 they should try to complete it in a quiet setting (e.g., while their child is sleeping). In 160 addition, there are more detailed instructions for completing the vocabulary checklist. 161 Unlike the traditional paper versions, instructions on how to properly choose responses are 162 provided both in written and pictorial form. The pictorial instructions (Figure 1) aim to 163 further increase caregivers' understanding of how to complete the checklist. For example, 164 these instructions clarify that the child's understanding of a word requires them to have 165 some understanding of the object that the word refers to or some aspect of the word's 166 meaning. In addition, caregivers are reassured that "child like" forms (e.g., "raff" for 167 "giraffe") or family- or dialect-specific forms (e.g., "nana" for "grandma" are acceptable). Lastly, caregivers are reminded that the child should be able to produce the words "on their own" and that imitations are not acceptable. These general "rules of thumb" for completing the form should be familiar to researchers who are distributing the forms to 171 parents so they can field any questions that may arise. While this is not possible for certain 172 use-cases (e.g., collecting data via Facebook), these instructions should ideally also be 173 reviewed either in writing (e.g., via email) or verbally (e.g., over the phone), so that these 174 pictured instructions serve merely as a reminder to caregivers when completing the form. 175

Completing the instrument. The majority of the participant's time in the study is spent completing the main sections of the instruments. As shown in Figure 2, on the American English Words and Gestures form, the vocabulary checklist portion of the form (396 items) asks parents to indicate whether their child can "understand" or "understand and say" each word. Gesture communication and other early milestones are also assessed. In the American English Words and Sentences form, the vocabulary checklist (680 items) only asks parents to indicate which words their child "says". Additional items assess children's production of their three longest sentences, as well as morphological and

184 syntactic development more broadly.

At the completion of the form, a graph is displayed illustrating the proportion of 185 words from each semantic category that the child currently produces or understands. In 186 addition, data from the norming studies are used to estimate the "hardest" (i.e., most 187 advanced) word that the child currently understands or produces. This feedback to parents 188 is intended to provide parents with a fun "thank you" and is intentionally not designed to 189 provide specific feedback about their child's progress. The closing page also reminds 190 parents that their participation does not constitute a clinical evaluation and that they 191 should contact their pediatrician or primary care physician if they have any concerns about 192 their child's development. 193

## 194 $Researcher\ Interface$

One of the main goals of Web-CDI is to facilitate wide distribution of the platform
to the child language research community. To that end, researchers are required to contact
a member of the CDI Advisory Board to register an account on Web-CDI, from which they
can create studies to distribute to participants. Note that we ask that researchers allow
fully anonymised data to be shared with us, so that it can be added to Wordbank
(http://wordbank.stanford.edu/). However, there is an opt-out option if researchers do not
wish to share their data.

A study in the context of the Web-CDI system is a set of individual administrations created by a researcher that share certain specifications. Table 1 gives an overview of the customizable features that are available at the study level in Web-CDI. These features are set when creating a study for the first time in Web-CDI using the "Create Study" tool, and most of the features can be updated continuously during data collection using the "Update Study" tool. While some of these features are only particularly relevant to specific use cases (e.g., longitudinal research and social media data collection, outline below), others are relevant to all researchers using Web-CDI.

There are currently several forms available for distribution on Web-CDI, including
multiple versions of the English WG and WS forms and forms in other languages (see
Cross-linguistic research below). When creating a study, researchers choose one of the
forms that they would like to distribute to participants; only one can be used in a given
study. Researchers who wish to send multiple forms to participants simultaneously (e.g.,
those conducting multilingual research) should create multiple studies, each with a single
instrument associated with it.

Researchers can download participant data in two formats. Both formatting options
output a comma-separated values file with one row per participant; the full data option
includes participant-by-item responses, and allows researchers to explore item-level trends,
while the summary data option omits item-level data and only provides summary scores
(e.g., total number of words understood/produced, percentile scores by age and gender).

Below, we outline several possible use cases of Web-CDI, as well the features which may facilitate them from a researcher's perspective.

Individual recruitment. One possible workflow using Web-CDI is to send unique 224 study URLs to individual participants. Researchers do so by entering numerical participant IDs or by auto-generating a specified quantity of participant IDs, each with its own unique study URL, using the "Add Participants" tool in the researcher dashboard. New 227 participants can be added on a continual basis so that researchers can adjust the sample 228 size of their study during data collection. Unique links generated for individual 229 participants expire, by default, 14 days after creation, though the amount of days before 230 link expiration is adjustable, which may be an important consideration for some 231 researchers depending on their participant populations and specific project timelines. This 232 workflow is most suitable for studies which pair the CDI with other measures, or when 233 researchers contact specific participants from an existing database. 234

Longitudinal studies. Web-CDI also facilitates longitudinal study designs in which

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each participant completes multiple administrations. Researchers wishing to design 236 longitudinal studies can do so by entering a list of meaningful participant IDs using the 237 "Add Participants" tool in the researcher dashboard. If a certain participant ID is added 238 multiple times, Web-CDI will create multiple unique study URLs in the study dashboard 239 that have the same specified ID. In addition, when creating studies, researchers can select 240 whether they would like the demographics information, vocabulary checklist, or no sections 241 at all to be prefilled when a participant fills out a repeat administration of the instrument. 242 Unless researchers are interested in cumulative vocabulary counts, it is strongly 243 recommended that they do not use the option to pre-fill the vocabulary checklist portion of 244 the instrument in longitudinal administrations as parents should complete the instrument 245 at each time point independently. 246

Social media and survey vendors. Web-CDI contains several features designed to 247 facilitate data collection from social media recruitment or through third-party 248 crowd-sourcing applications and vendors (e.g., Amazon Mechanical Turk, Prolific). First, 249 rather than creating unique survey links for each participant, researchers can also use a 250 single, anonymous link. When a participant clicks the anonymous link, a new 251 administration with a unique subject ID is created in the study dashboard. Additionally, Web-CDI studies have several customizable features that are geared towards anonymous 253 online data collection. For example, researchers can adjust the minimum amount of time a participant must take to fill out the survey before they are able to submit; with a longer 255 minimum time to completion, researchers can encourage a more thorough completion of 256 the survey. Researchers can also ask participants to verify that their information is 257 accurate by checking a box at the end of the survey, and can opt to include certain 258 demographic questions at both the beginning and end of the survey, using response 259 consistency on these redundant items as a check of data quality. 260

Paid participation. If researchers choose to compensate participants directly through the Web-CDI interface, Web-CDI has built-in functionality to distribute

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redeemable gift codes when a participant reaches the end of the survey. Web-CDI contains
several features to facilitate integration with third-party crowdsourcing applications and
survey vendors should they choose to handle participant compensation through another
platform. For example, when creating studies, researchers can enter a URL to redirect
participants to when they reach the end of the survey. Researchers using the behavioral
research platform Prolific can configure their study to collect participants' unique Prolific
IDs and pre-fill them in the survey.

Cross-linguistic research. Web-CDI forms are currently available in English (U.S.
American and Canadian), Spanish, French (Quebecois), Hebrew, Dutch and Korean. We
are looking to add more language forms to the tool as the paper version of the forms has
been adapted into more than 100 different languages and further ongoing adaptations have
been approved by the MB-CDI board (http://mb-cdi.stanford.edu/adaptations).

# $^{_{75}}$ $Administrator\ Interface$

System administrators who oversee the development and usage of Web-CDI can log 276 into a specific interface provided by Django, the open-source web framework used to 277 develop Web-CDI. This Django interface gives access to the entire dataset collected by all 278 participating researchers, as well as all of the instrument forms that can be distributed to 279 participants. Administrators approve new researchers' requests to create accounts on 280 Web-CDI, and give individual researchers access to specific instrument forms. As an example, if a researcher wants to distribute the Spanish Words and Gestures form to participants, an administrator would need to give a researcher access to this instrument. Administrator privileges are limited to a handful of individuals, as most of the functionality accessible to administrators is not relevant to researchers using the survey for 285 their own purposes.

### $System \ Design$

Web-CDI is constructed using open-source software. All of the vocabulary data collected in Web-CDI are stored in a standard MySQL relational database, managed using Django and Python and hosted by Amazon Web Services. Individual researchers can download data from their studies through the researcher interface, and Web-CDI admins have access to the entire aggregate set of data from all studies run with Web-CDI. Website code is available in a GitHub repository https://github.com/langcog/web-cdi, where interested users can browse, make contributions and request technical fixes.

## 295 Data Privacy and GDPR Compliance

Web-CDI is designed from the ground up to be compliant with stringent human 296 subjects privacy protections across the world. First, for US users, we have designed Web-CDI based on the United States Department of Health and Human Services "Safe 298 Harbor" Standard for collecting protected health information as defined by the Health Insurance Portability and Accountability Act (HIPAA). In particular, participant names are never collected, birth dates are used to calculate age in months (with no decimal 301 information) and never stored, and geographic zip codes are trimmed to the first 3 digits. Because of the architecture of the site, even though participants enter zip codes and dates 303 of birth, these are never transmitted in full to the Web-CDI server. Since no identifying 304 information is being collected by the Web-CDI system, this feature ensures that Web-CDI 305 can be used by United States labs without a separate Institutional Review Board 306 agreement between users labs and Web-CDI (though of course researchers using the site 307 will need Institutional Review Board approval of their own research projects)<sup>2</sup>. 308

<sup>&</sup>lt;sup>2</sup> Issues of de-identification and re-identifiability are complex and ever changing. In particular, compliance with DHHS "safe harbor" standards does not in fact fully guarantee the impossibility of statistical re-identification in some cases and if potential users have questions, we encourage them to consult with an Institutional Review Board.

In the European Union (EU), research data collection and storage is governed by 309 the Generalized Data Protection Regulation (GDPR) and its local instantiation in the legal 310 system of the member states. Some of the questions on the demographic form contain 311 information that may be considered sensitive (e.g., information about children's 312 developmental disorders), and in some cases, the possibility of linking this sensitive 313 information to participant IDs exists, particularly when researchers draw on local databases 314 that contain full names and addresses for recruitment and contacting. As a result, issues 315 regarding GDPR compliance arise when transferring data outside the EU, namely to 316 Amazon Web Services servers housed in the United States. Following GDPR regulations, 317 these issues would make a data sharing agreement between data collectors and Amazon 318 Web Services necessary. In addition, all administrators who can access the collected data 319 would have to enter such an agreement, which needs updating whenever personnel changes occur. To overcome these hurdles and in consultation with data protection officers, we 321 opted to exploit the local technical expertise and infrastructure to set up a sister site housed on GDPR-compliant servers, currently available under webcdi.mpi.nl. This site is 323 updated synchronously with the main Web-CDI website to ensure a consistent user 324 experience and access to the latest features and improvements. This site has been used in 135 successful administrations so far and is the main data collection tool for an ongoing 326 norming study in the Netherlands. We are further actively advertising the option to use 327 the European site to other labs who are following GDPR guidelines and are planning 328 adaptations to multiple European languages, where copyright allows. 329

# 30 Current Usage and Data

One of the key benefits of Web-CDI use is that the system in effect becomes a
centralized repository for standardized administrations of the CDI, contributing
anonymized data (again, on an opt-out basis) to future research and norming efforts. In
this section, we provide some preliminary analyses of the American English Web-CDI,

demonstrating the potential of the Web-CDI system to provide a distributed platform for gathering large CDI datasets.

At time of writing, researchers from 15 universities in the United States have 337 collected over 5,000 administrations of the American English CDI using Web-CDI since it 338 was launched in late 2017, with 2,868 administrations of the WG form and 2,868 339 administrations of the WS form. We excluded participants from the subsequent analyses based on a set of stringent criteria intended for the creation of future normative datasets. We excluded participants if it was not their first administration of the survey; if they were born prematurely or had a birthweight under 5.5 lbs (< 2.5 kg); reported more than 16 343 hours of exposure to a language other than English per week on average (amounting to 344 >10\% exposure to English); had serious vision impairments, hearing deficits or other developmental disorders or medical issues<sup>3</sup>; completed the survey unrealistically quickly 346 (defined here as in fewer than 8.5 minutes); or were outside of the correct age range for the 347 survey. The exclusion criteria we used differed from the most recent norming study of the 348 WG measure in the United Kingdom (Alcock, Meints, Rowland, & others, 2020) in order 349 to comply more closely with the criteria used in Fenson et al. (2007), who used more 350 stringent criteria to establish vocabulary norms that reflect typically developing children's 351 vocabulary trajectories. A complete breakdown of the number of participants excluded on 352 each criterion is in Table 2. Of the completed WG forms, 1,292 were excluded, leading to a 353 final WG sample size of 1,576 administrations, and 920 WS administrations were excluded, 354 leading to a final WS sample size of 1,948. 355

## 356 Demographic distribution and exclusions

Figure 2 shows the distribution of participant ethnicities as compared with
previously reported numbers in a large scale norming study of the paper-based CDI form

<sup>&</sup>lt;sup>3</sup> Exclusions on the basis of child health were decided on a case-by-case basis by author V.M. in consultation with Philip Dale, Donna Thal and Larry Fenson.

by Fenson et al. (2007). White participants still comprised nearly three quarters of the 359 Web-CDI sample, while a higher proportion of participants report mixed ethnic 360 identification as compared to the 2007 norms. Few participants identified as 361 Hispanic/Latino: 6.5% of WG participants and 5.1% of WS participants reported Hispanic 362 of Latino heritage. The low percentage of Hispanic/Latino participants was due in part to 363 our exclusion of children with substantial exposure to languages other than English. 364 Participants' educational attainment level was similarly skewed. Over 80% of children in 365 the Web-CDI sample came from families with college-educated mothers compared to 43% from the same group in the 2007 norms (Figure 3). Furthermore, less than 1 percent of 367 participants in our families report a maternal education level less than a high school 368 degree, compared to 7\% from the same group in the 2007 norms. The overrepresentation of 369 white Americans with high levels of education attainment in this sample points to a general challenge encountered in vocabulary development research, which we return to 371 when we detail our efforts to recruit more diverse participants.

#### Results to date

Although the CDI instruments include survey items intended to measure constructs 374 other than vocabulary size, such as gesture, sentence production and grammar, we focus 375 exclusively on the vocabulary measures here. Across both the WG and WS measures, our 376 current Web-CDI sample shows greater reported vocabulary comprehension and production 377 for older children. Moreover, data from both measures replicate a subtle but reliable 378 pattern found in Fenson et al. (1994), (2007) and Feldman et al. (2000) such that female 379 children tend to have slightly larger vocabulary scores than male children across the period 380 of childhood assessed in the CDI forms, though this difference does not appear until the 381 second year of life (Figure 3). 382

On the WG form, respondents' reports of children's vocabulary comprehension and production both increased with children's age (Figure 4). We replicate overall patterns

found by Feldman et al. (2000) in that, on both the "Words Understood" and "Words 385 Produced" measures, vocabulary scores were slightly negatively correlated with primary 386 caregivers' education level, such that those parents without any college education reported 387 higher vocabulary scores on both scales. A linear regression model with robust standard 388 errors predicting comprehension scores with children's age and primary caregivers' 389 education level (binned into categories of "High school diploma or less", "Some college 390 education" and "College diploma or more<sup>4</sup>") as predictors shows main effects of both age 391  $(\beta = 19.89, p < 0.001)$  and caregiver primary education  $(\beta_{highschool} = 29.59, p = 0.01)$ . 392 Similarly, a linear regression model with robust standard errors predicting production 393 scores by children's age and primary caregivers' education level shows main effects of age 394  $(\beta = 7.82, p < 0.001)$  and caregiver primary education  $(\beta_{highschool} = 28.86, p = 0.002)$ . 395 The pattern of results seen in the WG sample is consistent with prior findings 396 indicating that respondents with lower levels of education attainment report higher 397 vocabulary comprehension and production on the CDI-WG form (Feldman et al., 2000; 398 Fenson et al., 1994). Although caregivers with lower levels of education attainment report 399 400

higher mean levels of vocabulary production and comprehension, median vocabulary scores
show no clear pattern of difference across primary caregiver education levels (Figure 5).
This discrepancy between the regression effects and a group-median analysis suggests that
the regression effects described previously are driven in part by differential interpretation of
the survey items, such that a few lower-SES caregivers are more liberal in reporting their
children's productive and comprehensive vocabularies, especially for the youngest children,
driving up the mean scores for this demographic group.

Vocabulary production scores on the WS form show the expected pattern of increase with children's age in months; in addition, scores replicate the trend reported in Feldman et al. (2000) such that maternal education is positively associated with children's

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<sup>&</sup>lt;sup>4</sup> "High school diploma" or less corresponds to 12 or fewer years of education; "Some college" corresponds to 13 - 15 years of education; "College diploma or more" refers to 16 or more years of education.

reported vocabulary size (Figure 6). Because representation of caregivers without a high 410 school diploma is scarce (N = 18 out of a sample of 1,948), interpretation of the data from 411 this group is constrained. Nevertheless, as shown in Figure 6, a small but clear positive 412 association between maternal education and vocabulary score exists such that 413 college-educated caregivers report higher vocabulary scores than those of any other 414 education level. The implications from these data converge with previous findings which 415 indicate that parental education levels, often used as a metric of a family's socioeconomic 416 status, are correlated with children's vocabulary size through early childhood. 417

The data discussed above have stemmed from efforts by many researchers across the
United States whose motivations for using the Web-CDI vary. As a result, they reproduce
many of the biases of standard US convenience samples. In the next section, we describe in
more detail our recent efforts to use the Web-CDI to collect vocabulary development data
from traditionally underrepresented participant populations in the United States,
attempting to counteract these trends.

### Using Web-CDI to Collect Data from Diverse U.S.-based Communities

As discussed above, much past research on children's language development in the 425 US has tended to draw upon convenience samples of participants located near major U.S. 426 universities. Despite the large sample sizes we collected in the previous section, our current 427 dataset from Web-CDI is, if anything, even more biased towards highly-educated White 428 families than previous datasets. How can we recruit more diverse samples to remedy this issue? Here we discuss some potential routes forward. Importantly, we focus on collecting data from monolingual English-speaking families. While understanding the performance of standard measurement tools like the CDI among multilinguals is of immense import to the field of vocabulary development research (Gonzalez et al., in prep; Floccia et al., 2018; De 433 Houser, 2019), we focused here only on monolingual development, because collecting data 434 from multilingual populations introduces additional methodological considerations 435

(e.g. how to measure exposures in each language) which we do not address here.

#### $_{7}$ Online Data Collection

Online recruitment methods, such as finding participants on platforms such as
Amazon Mechanical Turk, Facebook and Prolific, represent one possible route towards
assembling a large, diverse sample to take the Web-CDI. However, these recruitment
methods are largely untested with parent report measures of child language development.
In a series of data collection efforts, we used Web-CDI as a tool to explore these different
channels of recruitment.

In our first phase of data collection, we ran advertisements on Facebook which were aimed at non-white families based on users' geographic locations (e.g., targeting cities 445 which have a higher than average representation of African Americans) or other profile 446 features (e.g., ethnic identification, interest in parenthood-related topics). Advertisements 447 consisted of an image of a child and a caption informing Facebook users of an opportunity 448 to fill out a survey on their child's language development and receive an Amazon gift card 449 (Figure 7). Upon clicking the advertisement, participants were redirected to a unique 450 administration of the Web-CDI, and they received \$5 upon completing the survey. This 451 open-ended approach to recruitment offered several advantages, namely that a wide variety 452 of potential participants from specific demographic backgrounds can be reached on 453 Facebook. However, we also received many incomplete or otherwise unusable survey 454 administrations, either from Facebook users who click the link and decide not to 455 participate, or those who completed the survey in an unrealistically short period of time (over half of all completed administrations, Table 3). 457

In the second phase of our data collection efforts, we used the crowdsourcing survey vendor Prolific (http://prolific.co) in the hopes that some of the challenges encountered with Facebook recruitment would be addressed. Prolific allows researchers to create studies and post them to individuals who are in the platform's participant database, each of whom

is assigned a unique alphanumeric "Prolific ID". Importantly, Prolific maintains detailed 462 demographic information about participants, allowing researchers to specify whom they 463 would like to complete their studies. Prolific further has a built-in compensation 464 infrastructure that handles monetary payments to participants, eliminating the need to 465 disburse gift cards through Web-CDI. In the particular case of Web-CDI, the demographic 466 information needed to determine whether an individual is eligible to complete the survey 467 (e.g., has a child in the correct age range, lives in a monolingual household, etc.) is more 468 specific than the information that Prolific collects about their participant base. We 460 therefore used a brief pre-screening questionnaire to generate a list of participants who 470 were eligible to participate, and subsequently advertised the Web-CDI survey to those 471 participants. Given that we were interested only in reaching participants in the United 472 States who were not white or who did not have a college diploma, our data collection efforts only yielded a sample that was small (N = 71) but much more thoroughly screened 474 than that which we could obtain on Facebook.

Across both phases, we used the same exclusion criteria as in the full Web-CDI 476 sample to screen participants. A complete tally of all excluded participants is shown in Table 3. In both the WG and WS surveys, exclusions rates were high, amounting to 58% of participants who completed the survey. The high exclusions rates were notably driven by 479 an accumulation of survey administrations which participants completed very quickly (in these analyses, defined as a completion taking less than 8.5 minutes). Many of the survey 481 administrations excluded for fast completion had missing demographic information 482 reported: Among WG participants excluded for too-fast completions, 93% did not report ethnicity, and among WS participants excluded for the same reason, 2% did not report 484 ethnicity. After exclusions, full sample size was N = 115 WG completions and N = 126485 completions. 486

The results from our recent data collection efforts focused on lower-SES, non-white participants show overall similar patterns to the full Web-CDI sample in several regards.

Word production scores from both the WG and WS administrations reflect growing
productive vocabulary across the second and third years, with a very small gender effect
such that female children's vocabularies grow at a slightly faster rate than males' (Figure
8). The relationship between caregivers' reported levels of maternal education and child's
vocabulary score is not as clear as it is in the full Web-CDI sample (Figure 9); however,
children of college-educated parents show slightly faster vocabulary growth than do
children of parents without any college degree.

Importantly, recent data collection efforts showed a substantial improvement in 496 reaching non-white or less highly-educated participants. After exclusions, the Web-CDI data we collected through Facebook and Prolific have a higher proportion of non-white 498 participants than the overall Web-CDI sample and the norms established by Fenson et al. 499 (2007) (Figure 10). Black participants in particular showed a marked increase in 500 representation, from 10.5% in the 2007 norms to 30.9% in the recent sample, while the 501 proportion of white participants decreased from 73.3% in the 2007 norms to 50.2% in the 502 recent sample. Representation on the basis of families' reported primary caregiver 503 education also improved (Figure 10). Participants with only a high school diploma 504 accounted for 31.5% of the recent sample as compared to 23.8% in the 2007 norms, and 505 representation of those with a college diploma or more education decreased from 43.8% in 506 the 2007 norms to 36.5% in the recent sample. 507

#### 508 Discussion

Taken together, these recent results indicate that Web-CDI could be a promising
avenue through which to collect vocabulary development data in non-white, lower-SES
communities when paired with online recruitment methods that yield legitimate,
representative participant samples. These data do, however, convey clear limitations of our
approach. Perhaps most conspicuously, more than half of completed administrations in this
sample had to be excluded, in many cases because the information provided by participants

appeared rushed or incomplete: over 40% of administrations were completed in fewer than 515 8.5 minutes, and of these quick completions, well over 90% were missing demographic 516 information that is rarely missing in other administrations of the form. Determining the 517 precise reasons for the high exclusion rate, and how (if at all) this may skew data reflecting 518 demographic trends in vocabulary development, requires a more thorough assessment of 519 who is submitting hastily-completed forms, which is beyond the scope of the current study. 520 However, all respondents who got to the end of the form were compensated regardless of 521 how thoroughly they completed it, creating the possibility that some participants who 522 clicked the anonymous link may not have been members of the population of interest, but 523 rather were other individuals motivated by the ending compensation. General scarcity of 524 demographic information on these individuals prevents us from making speculative claims 525 about which groups constituted the bulk of participant exclusions.

Additionally, the exclusion rates described previously only provide information on those participants who did, at some point, submit a completed form, but many individuals 528 clicked the advertisement link and did not subsequently continue on to complete the form. 529 Without an in-depth exploration of who is clicking the link and why they might choose not 530 to continue, we cannot draw conclusions about the representativeness of the current sample with regards to the communities we would like to include in our research. As such, a more thorough understanding of how users from different communities respond to various recruitment and sampling methods is needed in future work in order to draw conclusions about demographic trends above and beyond those already established in the literature. 535

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In a similar vein, participants in this study were recruited through a targeted post 536 on social media, a technique that is considerably more anonymous than recruitment strategies which entail face-to-face or extended contact between researchers and community 538 members. For future research, it should be considered that online recruitment methods 539 may not be suitable for all communities, especially when researchers ask participants to 540 report potentially sensitive information about the health and development of their children 541

(even when such information is stored anonymously). Our goal here was to assess whether
general trends in past literature could be recovered using such an online strategy, but
future research should take into account that other more personal methods of recruitment,
such as those through more direct community out reach or liaison contacts, may improve
participants' experiences and their willingness to engage with the study.

Finally, a significant limitation of the current data collection process is that many people in the population of interest - particularly lower-income families - do not have reliable internet access. Having participants complete the Web-CDI on a mobile device may alleviate some of the issues caused by differential access to Wi-Fi, since the vast majority of American adults own a smartphone (Pew Research Center, 2018 - NEED TO 551 ADD MANUALLY TO REFS). Accordingly, improving Web-CDI's user experience on 552 mobile platforms will be an important step towards ensuring that caregivers across the 553 socioeconomic spectrum can easily complete the survey. For smartphone users on 554 pay-as-you-go plans, who may be reluctant to use data to complete a study, a possible 555 solution could be compensating participants for the amount of "internet time" they 556 incurred completing the form. 557

#### 58 Conclusions

In this paper, we presented Web-CDI, a comprehensive online interface for researchers to measure children's vocabulary by administering the MacArthur-Bates
Communicative Development Inventory family of parent-report instruments. Web-CDI provides a convenient researcher management interface, built-in data privacy protections, and a variety of features designed to make both longitudinal and social-media sampling easy. To date, over 4,000 valid administrations of the WG and WS forms have been collected on Web-CDI from more than a dozen researchers in the United States after applying strict exclusion criteria derived from previous norming studies) (Fenson et al., 2007, 1994).

Many research laboratories, not only in the United States but around the world, 568 collect vocabulary development data using the MacArthur-Bates CDI (Frank et al., 2021). 569 With traditional paper-based forms, combining insights from various research groups can 570 prove challenging, as each group may have slightly different ways of formatting and 571 managing data from CDI forms. By contrast, if all of these groups' data come to be stored 572 in a single repository with a consistent database structure, data from disparate sources can 573 easily be collated and analyzed in a uniform fashion. As such, a centralized repository such 574 as Web-CDI provides a streamlined data-aggregation pipeline that facilitates cross-lab 575 collaborations, multisite research projects and the curation of large datasets that provide 576 more power to characterize the vast individual differences present in children's vocabulary 577 development. 578

Beyond the goal of simply getting more data, we hope that Web-CDI can advance 579 efforts to expand the reach of vocabulary research past convenience samples into diverse 580 communities. A key question in the field of vocabulary development concerns the 581 mechanisms through which sociodemographic variables, such as race, ethnicity, income and 582 education are linked to group differences in vocabulary outcomes. Large, 583 population-representative samples of vocabulary development data are needed to understand these mechanisms, but most research to date (including the full sample of Web-CDI administrations) oversamples white participants and those with advanced levels of education. We explored the use of Web-CDI as part of a potential strategy to collect 587 data from non-white, lower-SES communities in two phases. Several overall patterns 588 emerged from the resulting data which we expected: vocabulary scores grew with age, 589 providing a basic validity check of the Web-CDI measure; females held a slight advantage 590 in word learning over males; and children of parents with a college education showed 591 slightly higher vocabulary scores. 592

Nonetheless, the insights from these data, while aligned with past norming studies, are necessarily constrained by several features of our method. First, exclusion rates among

data collected on Facebook were very high, well over 50%, mostly due to a large quantity of 595 hasty completions. Second, a rigorous evaluation of the population-representativeness of 596 those who were counted in the final sample was not feasible here. Given our findings, we 597 maintain that web-based data collection can capture useful information about vocabulary 598 development from diverse communities, but future research will need to examine which 599 sampling methods can yield accurate, population-representative data that can advance our 600 understanding of the link between sociodemographic variation and variation in language 601 outcomes. 602

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Table 1
Settings customizable by researchers when creating new studies to be run on the Web-CDI platform.

Study setting	Default	Notes
	value	
Study name	none	NA
Instrument	none	NA
Number of days before study	14	Must be between 1 and 28
expiration		days.
Measurement units for birth	Pounds and	Weight can also be measured
weight	ounces	in kilograms (kg).
Minimum time (minutes) a	6	NA
parent must take to complete		
the study		
Waiver of documentation	blank	Can be filled in by researchers
		to include a Waiver of
		Documentation for the
		participant to approve before
		proceeding to the experiment.
Pre-fill data for longitudinal	"No, do not	Researchers can choose to
participants?	populate any	pre-fill the background
	part of the	information and the
	form"	vocabulary checklist.

Table 1
Settings customizable by researchers when creating new studies to be run on the Web-CDI platform. (continued)

Study setting	Default	Notes
	value	
Would you like to pay subjects	No	If checked, researchers can
in the form of Amazon gift		enter gift codes to distribute
cards?		to participants once they have
		completed the survey.
Do you plan on collecting only	No	If checked, researchers can set
anonymous data in this study?		a limit for the maximum
(e.g., posting ads on social		number of participants, as well
media, mass emails, etc)		as select an option that asks
		participants to verify that the
		information entered is
		accurate.
Would you like to show	Yes	NA
participants graphs of their		
data after completion?		
Would you like participants to	No	NA
be able to share their		
Web-CDI results via		
Facebook?		

 $\begin{tabular}{ll} \textbf{Table 2} \\ Exclusions from full WebCDI sample \\ \end{tabular}$ 

Exclusion	WG	% of full	WS	% of full
	exclusions	WG sample	exclusions	WS sample
		excluded		excluded
Not first administration	163	5.68%	444	12.35%
Premature or low birthweight	37	1.29%	67	1.86%
Multilingual exposure	449	15.66%	492	13.69%
Illnesses/Vision/Hearing	191	6.66%	203	5.65%
Out of age range	88	3.07%	200	5.56%
Completed survey too quickly	363	12.66%	236	6.57%
System error in word	1	0.03%	4	0.11%
tabulation				
Total exclusions	1292	45%	1646	46%

 $\begin{tabular}{ll} \textbf{Table 3} \\ Exclusions from recent data collection using Facebook and Prolific. \end{tabular}$ 

Exclusion	WG	% of full	WS	% of full
	exclusions	WG sample	exclusions	WS sample
		excluded		excluded
Not first administration	0	0.00%	0	0.00%
Premature or low birthweight	7	2.53%	1	0.33%
Multilingual exposure	18	6.50%	23	7.62%
Illnesses/Vision/Hearing	4	1.44%	4	1.32%
Out of age range	1	0.36%	26	8.61%
Completed survey too quickly	132	47.65%	122	40.40%
System error in word	0	0.00%	0	0.00%
tabulation				
Total exclusions	162	58%	176	58%

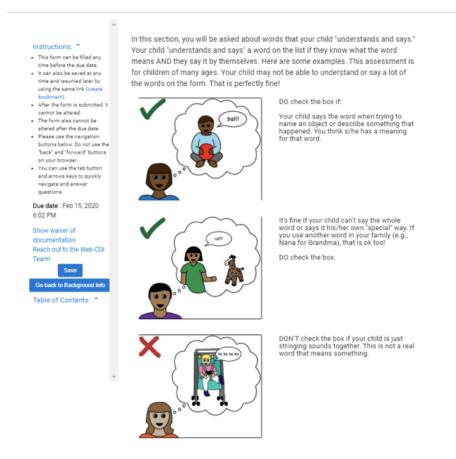


Figure 1

Pictorial instructions in the Web-CDI Words and Sentences instrument.

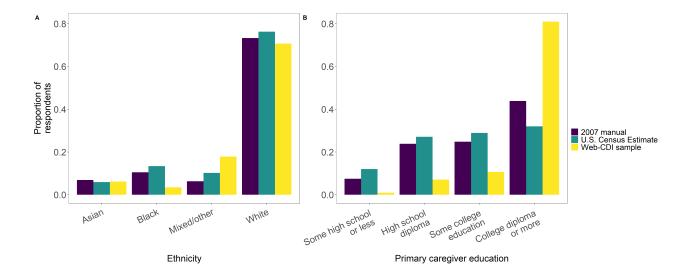
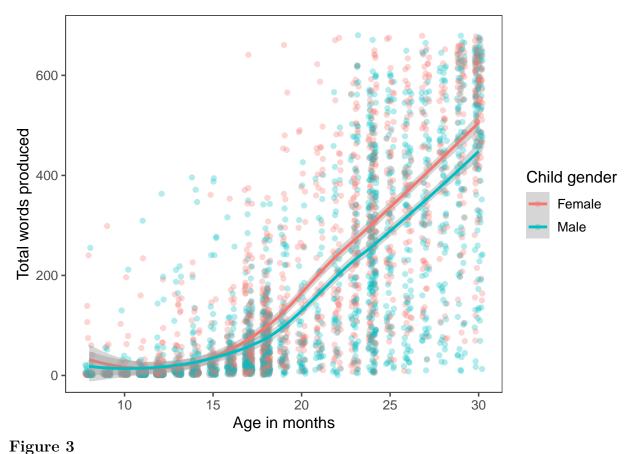


Figure 2

Proportion of respondents plotted by child race (A) and educational level of primary caregiver (B) from full Web-CDI sample to date (N=3,524), compared with norming sample demographics from Fenson (2007). Latinx participants can be of any race.



Individual children's vocabulary production scores from the entire Web-CDI sample plotted by children's age and gender (both WG and WS, N=3,513, with 1,674 girls). Line is a locally weighted regression with associated 95% confidence interval. Children with a different or no reported gender (N=11) are omitted here.

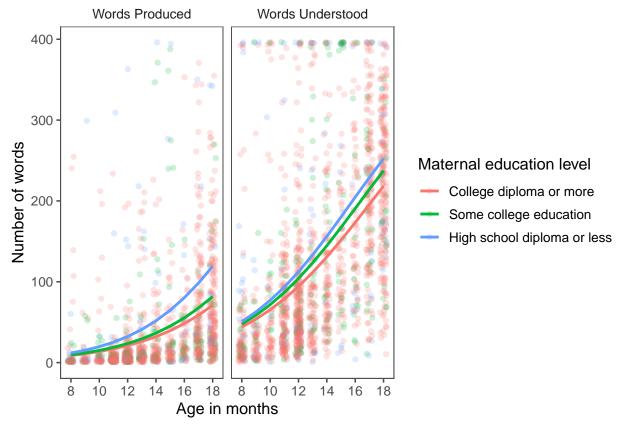


Figure 4

Individual children's word production (left panel) and comprehension (right panel) scores plotted by age and primary caregiver's level of education (binned into "High school diploma or less", "Some college education", and "College diploma or more") as reported in the sample of Words and Gestures Web-CDI administrations collected as of November 2020 (N=1,576). Curves show generalized linear models fits.

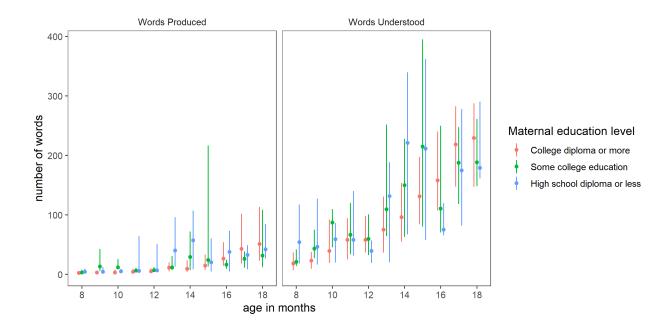


Figure 5

Median vocabulary production (left) and comprehension (right) scores by age on the WG form. Lines indicate span between first and third quartiles for each age.

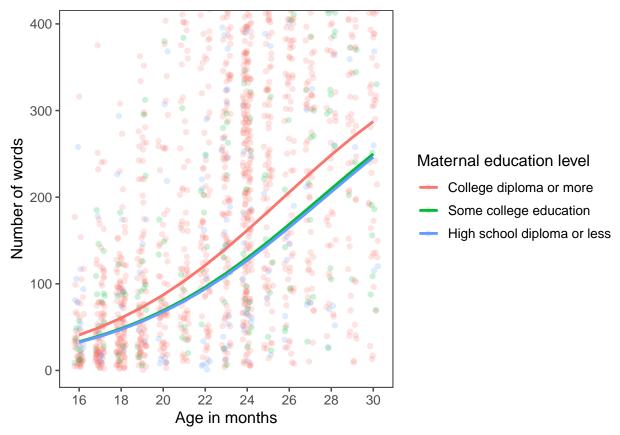


Figure 6

Individual children's vocabulary production scores plotted by children's age and maternal education level of primary caregiver education as reported in the sample of Words and Sentences Web-CDI administrations collected as of November 2020 (N=1,948). Lines are smoothed quantile regressions showing the estimated median vocabulary score within each education group at each age.



Calling all parents of 16 - 30 month old children! Do a 30 minute survey about your child's language ....See More

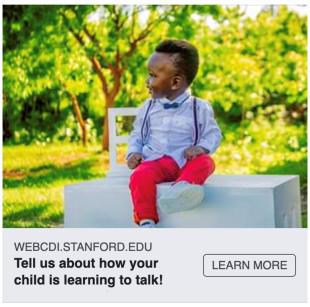
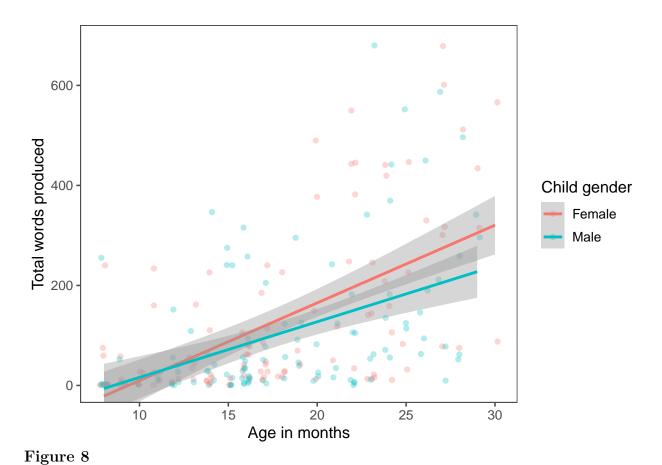
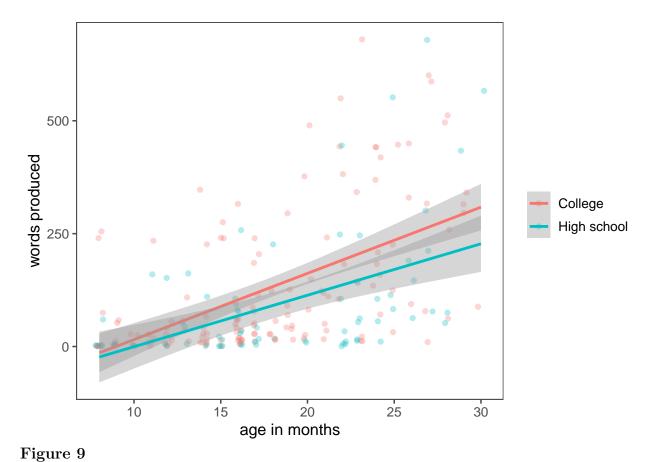


Figure 7

Example Facebook advertisement in Phase 1 of recent data collection.



Individual children's vocabulary production scores from the entire Web-CDI sample plotted by children's age and gender (both WG and WS, N=238, with 116 girls). Lines are best linear fits with associated 95% confidence intervals. Children with a different or no reported gender (N=3) are omitted here.



Individual children's vocabulary production scores plotted by age and level of primary caregiver education, binned into those with a high school diploma or less education and those with some college education or a college diploma (N=241). Lines show best linear fits and associated

95% confidence intervals.

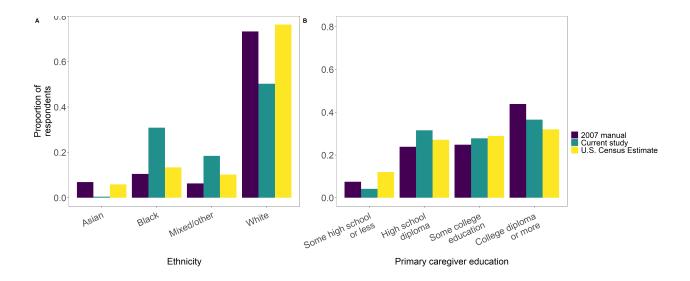


Figure 10

Proportion of respondents plotted by child race (A) and educational level of primary caregiver (B) from recent data collection efforts aimed towards oversampling non-white, lower-SES families (N=241), compared with norming sample demographics from Fenson (2007). Latinx participants can be of any race.