Web-CDI: A system for online administration of the MacArthur-Bates Communicative Development Inventories Benjamin deMayo¹, Danielle Kellier², Mika Braginsky³, Christina Bergmann⁴, Cielke

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Abstract

Understanding the mechanisms that 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 collect such data, but the traditional paper-based forms make the curation of large datasets logistically challenging. Many CDI 15 datasets are thus gathered using convenience samples, often recruited from communities in 16 proximity to major research institutions. Here, we introduce Web-CDI, a web-based tool 17 which allows researchers to collect CDI data online. Web-CDI contains functionality to 18 collect and manage longitudinal data, share links, and download standardized vocabulary 19 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: 21 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 23 educational attainment report slightly higher vocabulary production scores. We also report results from an effort to oversample non-white, lower-SES participants via online recruitment 25 (N = 241). These data showed similar demographic trends to the full sample but this effort 26 recruited in a high exclusion rate. We conclude by discussing implications and challenges for 27 the collection of large, population-representative datasets. 28

29 Keywords: vocabulary development, parent report, socioeconomic status

Word count: X

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Children vary tremendously in their vocabulary development (Fenson et al., 1994;
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, which was 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.

The Importance of Parent Report Data

Gaining empirical traction on variation in children's early language requires reliable and valid methods for measuring language abilities, especially in early childhood (8 to 30 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). 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 structure.
Together, these instruments allow for a comprehensive picture of milestones that characterize
language development in early childhood.

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 (e.g., Bates & Goodman, 2001; Bornstein & Putnick,
2012; see also, Frank et al., 2021). The popularity of CDI instruments has remained strong
over the years, leading to 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 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 (Fenson et al., 2007).
Representation in these norming samples is generally restricted to families living on the US
east and west coasts. 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 92 the MacArthur-Bates CDIs in order to address some of the limitations of the standard paper 93 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 95 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 97 uses an online administration format. 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, & Pyers, 2020). The innovation of Web-CDI is to provide a 100 comprehensive researcher management interface for the administration of a wide range of 101 CDI forms, allowing researchers to manage longitudinal administrations, download 102 standardized scores, and share links easily, all while satisfying strong guarantees regarding 103 privacy and anonymity. Moreover, a key benefit of a unified data collection and storage 104 system such as Web-CDI is that data from disparate sources are combined into a single 105 repository, substantially reducing the overhead efforts associated with bringing together data 106 collected using paper forms by researchers across the world. 107

Introducing Web-CDI

Web-CDI is a web-based platform for CDI administration and management. 109 Web-CDI allows researchers to communicate with families by sharing URLs via email or 110 social media, facilitating access to families in areas distant from an academic institution and 111 eliminating costly mailings and laboratory visits. Web-CDI also standardizes electronic 112 administration and scoring of CDI forms across labs and institutions, facilitating aggregation 113 of CDI data for later reuse and comparison across administrations by different labs. Indeed, 114 users of Web-CDI grant the CDI Advisory Board to access and analyze the resulting data on 115 an opt-out basis, providing a path towards continual improvement of CDI instruments. Since 116 2018, more than 3500 CDIs have been collected by 15 research groups throughout the US 117 who are using Web-CDI, demonstrating the potential for large-scale data collection and 118 aggregation. 119

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.

Participant interface

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Participants can complete the Web-CDI on a variety of devices, including personal 124 computers and tablets. Web-CDI can be administered on a smartphone, although the 125 experience is not as ideal for the user due to the length of the survey. (As Web-CDI moves in 126 the future to incorporate more short forms and adaptive forms, smartphone responsive design will become a priority). When a participant clicks a URL shared by a researcher, they 128 are directed to a website displaying their own personal administration of the Web-CDI, regardless of whether the link was participant-specific or general-purpose. In some cases, 130 they may be asked to read and accept a waiver of consent documentation, depending on 131 whether the researcher has chosen to use that feature (see also Researcher Interface below). 132

Demographics. The parent is next asked to provide demographic information about 133 their family and any health conditions that might impact their child's vocabulary 134 development. The specific demographic questions asked of participants can be adjusted to 135 vary between different versions of the form, allowing researchers to tailor the demographic 136 questions to local norms. For example, the Dutch CDI omits questions about participant 137 ethnicity, since census data in the Netherlands does not include ethnicity. Researchers can 138 customize the presentation of these demographic questions in three ways. First, they can 139 elect to show all of the demographics items on the landing page or to present the majority of 140 these questions at the end of the instrument. This choice is provided because some pilot 141 work in the United Kingdom indicated that answering questions regarding personal health 142 information early in administration may deter participants from completing the instrument. 143 Second, certain demographic questions can be asked at both the beginning and the end of the form to serve as validity checks, providing a check that can be used to screen for hasty or illegitimate completions. Third, researchers can tailor the questions to the societal and cultural context of their participants (e.g., country-specific education level descriptors and income categories). 148

Instructions. After completing the first demographics page, participants are provided 149 with instructions that are appropriate for either the Words & Gestures or Words & 150 Sentences version (see Figure 1). At the top of the page are general instructions that inform 151 participants that they should expect the study to take at least 30 minutes and that they 152 should try to complete it in a quiet setting (e.g., while their child is sleeping). In addition, 153 there are more detailed instructions for completing the vocabulary checklist. Unlike the traditional paper versions, instructions on how to properly choose responses are provided 155 both in written and pictorial form. The pictorial instructions (Figure 1) aim to further increase caregivers' understanding of how to complete the checklist. For example, these 157 instructions clarify that the child's understanding of a word requires them to have some 158 understanding of the object that the word refers to or some aspect of the word's meaning. In 159

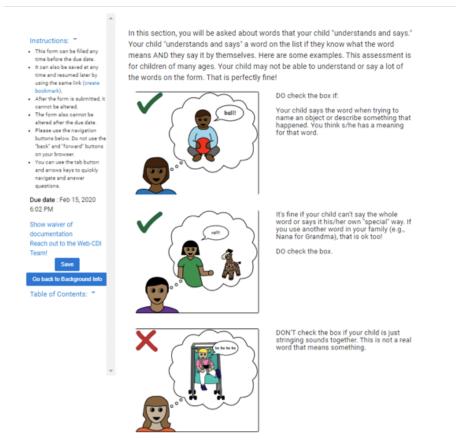


Figure 1

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

addition, caregivers are reassured that "child-like" forms (e.g., "raff" for "giraffe") or family-160 or dialect-specific forms (e.g., "nana" for "grandma" are acceptable). Lastly, caregivers are 161 reminded that the child should be able to produce the words "on their own" and that 162 imitations are not acceptable. These general "rules of thumb" for completing the form 163 should be familiar to researchers who are distributing the forms to parents so they can field 164 any questions that may arise. While this is not possible for certain use-cases (e.g., collecting 165 data via Facebook), these instructions should ideally also be reviewed either in writing (e.g., 166 via email) or verbally (e.g., over the phone), so that these pictured instructions serve merely 167 as a reminder to caregivers when completing the form.

Completing the instrument. The majority of the participant's time in the study is

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spent completing the main sections of the instruments. As shown in Figure 2, on the 170 American English Words and Gestures form, the vocabulary checklist portion of the form 171 (396 items) asks parents to indicate whether their child can "understand" or "understand 172 and sav" each word. Gesture communication and other early milestones are also assessed. In 173 the American English Words and Sentences form, the vocabulary checklist (680 items) only 174 asks parents to indicate which words their child "says". Additional items assess children's 175 production of their three longest sentences, as well as morphological and syntactic 176 development more broadly. All of these items are broken up across multiple screens for easier 177 navigation through the form. 178

At the completion of the form, a graph is displayed illustrating the proportion of 179 words from each semantic category that the child currently produces or understands. In 180 addition, data from the norming studies are used to estimate the "hardest" (i.e., most 181 advanced) word that the child currently understands or produces. This feedback to parents 182 is intended to provide parents with a fun "thank you" and is intentionally not designed to 183 provide specific feedback about their child's progress relative to other children or any 184 normative standard. The closing page also reminds parents that their participation does not 185 constitute a clinical evaluation and that they should contact their pediatrician or primary 186 care physician if they have any concerns about their child's development. 187

Researcher interface

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	

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

Study setting	Default value	Notes
Number of days before study	14	Must be between 1 and 28
expiration	11	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 vocabulary
	form"	checklist.
Would you like to pay subjects	No	If checked, researchers can
in the form of Amazon gift		enter gift codes to distribute to
cards?		participants once they have
		completed the survey.

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

Study setting	Default value	Notes
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?		

One of the main goals of Web-CDI is to provide a unified CDI platform to the child language research community. To that end, researchers request an account by contacting a member of the CDI Advisory Board. Once they have registered an account they can create studies to distribute to participants. One rationale for this personalized registration process is that we ask that researchers allow fully anonymized data from their participants to be shared with the CDI Advisory Board, so that it can be added to Wordbank (http://wordbank.stanford.edu/; Frank, Braginsky, Yurovsky, & Marchman, 2017) and shared with broader research community. However, there is an opt-out option if researchers

do not wish to share their data, making data contribution voluntary.

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A study in the context of the Web-CDI system is a set of individual administrations 198 created by a researcher that share certain specifications. Table 1 gives an overview of the 199 customizable features that are available at the study level in Web-CDI. These features are 200 set when creating a study for the first time in Web-CDI using the "Create Study" tool, and 201 most of the features can be updated continuously during data collection using the "Update 202 Study" tool. While some of these features are only particularly relevant to specific use cases 203 (e.g., longitudinal research and social media data collection, described below), others are 204 relevant to all researchers using Web-CDI. 205

There are currently several CDI forms available for distribution on Web-CDI, 206 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 208 forms that they would like to distribute to participants; only one can be used in a given 209 study. Researchers who wish to send multiple forms to participants simultaneously (e.g., 210 those conducting multilingual research) should create multiple studies, each with a single instrument associated with it. 212

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 218 may facilitate them from a researcher's perspective. 219

Individual recruitment. One possible workflow using Web-CDI is to send unique study 220 URLs to individual participants. Researchers do so by entering numerical participant IDs or 221

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 participants can 223 be added on a continual basis so that researchers can adjust the sample size of their study 224 during data collection. Unique links generated for individual participants expire, by default, 225 14 days after creation, though the amount of days before link expiration is adjustable, which 226 may be an important consideration for some researchers depending on their participant 227 populations and specific project timelines. This workflow is most suitable for studies which 228 pair the CDI with other measures, or when researchers contact specific participants from an 229 existing database. 230

Longitudinal studies. Web-CDI also facilitates longitudinal study designs in which 231 each participant completes multiple administrations. Researchers wishing to design 232 longitudinal studies can do so by entering a list of meaningful participant IDs using the 233 "Add Participants" tool in the researcher dashboard. If a certain participant ID is added 234 multiple times, Web-CDI will create multiple unique study URLs in the study dashboard 235 that have the same specified ID. In addition, when creating studies, researchers can select 236 whether they would like the demographics information, vocabulary checklist, or no sections 237 at all to be prefilled when a participant fills out a repeat administration of the instrument. 238 Unless researchers are interested in cumulative vocabulary counts, it is strongly 230 recommended that they do not use the option to pre-fill the vocabulary checklist portion of 240 the instrument in longitudinal administrations as parents should complete the instrument at 241 each time point independently. 242

Social media and survey vendors. Web-CDI contains several features designed to
facilitate data collection from social media recruitment or through third-party
crowd-sourcing applications and vendors (e.g., Amazon Mechanical Turk, Prolific). First,
rather than creating unique survey links for each participant, researchers can also use a
single, anonymous link. When a participant clicks the anonymous link, a new 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 online data collection. 240 For example, researchers can adjust the minimum amount of time a participant must take to 250 fill out the survey before they are able to submit; with a longer minimum time to completion, 251 researchers can encourage a more thorough completion of the survey. Researchers can also 252 ask participants to verify that their information is accurate by checking a box at the end of 253 the survey, and can opt to include certain demographic questions at both the beginning and 254 end of the survey, using response consistency on these redundant items as a check of data 255 quality.

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

System Design

Web-CDI is constructed using open-source software. All of the vocabulary data 272 collected in Web-CDI are stored in a standard MySQL relational database, managed using 273 Django and Python and hosted either by Amazon Web Services or by an Eropean Union 274 compliant server (see below). Individual researchers can download data from their studies 275 through the researcher interface, and Web-CDI admins have access to the entire aggregate 276 set of data from all studies run with Web-CDI. Website code is available in a GitHub 277 repository https://github.com/langcog/web-cdi, where interested users can browse, make 278 contributions, and request technical fixes. 279

280 Data Privacy and GDPR Compliance

Web-CDI is designed to be compliant with stringent human subjects privacy 281 protections across the world. First, for US users, we have designed Web-CDI based on the 282 United States Department of Health and Human Services "Safe Harbor" Standard for 283 collecting protected health information as defined by the Health Insurance Portability and 284 Accountability Act (HIPAA). In particular, participant names are never collected, birth 285 dates are used to calculate age in months (with no decimal information) and never stored, 286 and geographic zip codes are trimmed to the first 3 digits. Because of the architecture of the 287 site, even though participants enter zip codes and dates of birth, these are never transmitted 288 in full to the Web-CDI server. Since no identifying information is being collected by the 280 Web-CDI system, this feature ensures that Web-CDI can be used by United States labs 290 without a separate Institutional Review Board agreement between users labs and Web-CDI 291 (though of course researchers using the site will need Institutional Review Board approval of 292 their own research projects).¹ 293

¹ 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

In the European Union (EU), research data collection and storage is governed by the 294 Generalized Data Protection Regulation (GDPR) and its local instantiation in the legal 295 system of the member states. Some of the questions on the demographic form contain 296 information that may be considered sensitive (e.g., information about children's 297 developmental disorders), and in some cases, the possibility of linking this sensitive 298 information to participant IDs exists, particularly when researchers draw on local databases 290 that contain full names and addresses for recruitment and contacting. As a result, issues 300 regarding GDPR compliance arise when transferring data outside the EU, namely to 301 Amazon Web Services servers housed in the United States. Following GDPR regulations, 302 these issues would make a data sharing agreement between data collectors and Amazon Web 303 Services necessary. In addition, all administrators who can access the collected data would 304 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 opted to exploit the local technical expertise and infrastructure to set up a sister site housed on GDPR-compliant servers, currently available at http://webcdi.mpi.nl. This site is updated 308 synchronously with the main Web-CDI website to ensure a consistent user experience and 300 access to the latest features and improvements. This site has been used in 135 successful 310 administrations so far and is the main data collection tool for an ongoing norming study in 311 the Netherlands. We are further actively advertising the option to use the European site to 312 other labs who are following GDPR guidelines and are planning adaptations to multiple 313 European languages, where copyright allows. 314

Current Web-CDI Usage

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

Institutional Review Board.

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Table 2

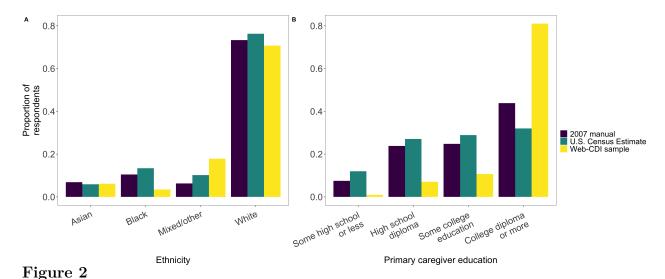
Exclusions from full WebCDI sample

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%

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 collected over 5,000 administrations of the American English CDI using Web-CDI since it was launched in late 2017, with 2,868 administrations of the WG form and 2,868 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 hours of exposure to a language other than English per week on average (amounting to >10%

exposure to English); had serious vision impairments, hearing deficits or other developmental 330 disorders or medical issues²; completed the survey unrealistically quickly (defined here as in 331 fewer than 8.5 minutes; see below for some justification of this number); or were outside of 332 the correct age range for the survey. The exclusion criteria we used were similar to those 333 used in Fenson et al. (2007), who adopted stringent criteria to establish vocabulary norms 334 that reflect typically developing children's vocabulary trajectories. A complete breakdown of 335 the number of participants excluded on each criterion is in Table 2. Of the completed WG 336 forms, 1,292 were excluded, leading to a final WG sample size of 1,576 administrations, and 337 920 WS administrations were excluded, leading to a final WS sample size of 1,948. 338



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.

Demographic distribution and exclusions

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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 by Fenson et

² 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.

al. (2007). White participants still comprised nearly three quarters of the Web-CDI sample, 342 while a higher proportion of participants report mixed ethnic identification as compared to 343 the 2007 norms. Few participants identified as Hispanic/Latino: 6.5\% of WG participants 344 and 5.1% of WS participants reported Hispanic of Latino heritage. The low percentage of 345 Hispanic/Latino participants was due in part to our exclusion of children with substantial 346 exposure to languages other than English. Participants' educational attainment level was 347 similarly skewed. Over 80% of children in the Web-CDI sample came from families with 348 college-educated mothers compared to 43% from the same group in the 2007 norms (Figure 3). Furthermore, less than 1 percent of participants in our families report a maternal 350 education level less than a high school degree, compared to 7% from the same group in the 351 2007 norms. The overrepresentation of white Americans with high levels of education 352 attainment in this sample points to a general challenge encountered in vocabulary development research, which we return to when we detail our efforts to recruit more diverse participants. 355

Results

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Although the CDI instruments include survey items intended to measure constructs other than vocabulary size, such as gesture, sentence production and grammar, we focus 358 exclusively on the vocabulary measures here. Across both the WG and WS measures, our 359 current Web-CDI sample shows greater reported vocabulary comprehension and production 360 for older children. Moreover, data from both measures replicate a subtle but reliable pattern such that female children tend to have slightly larger vocabulary scores than male children across the period of childhood assessed in the CDI forms (Frank et al., 2021), though in 363 these data this difference does not appear until around 18 months (Figure 3).

On the WG form, respondents' reports of children's vocabulary comprehension and 365 production both increased with children's age (Figure 4). We replicate overall patterns found 366 by Feldman et al. (2000) in that, on both the "Words Understood" and "Words Produced" 367

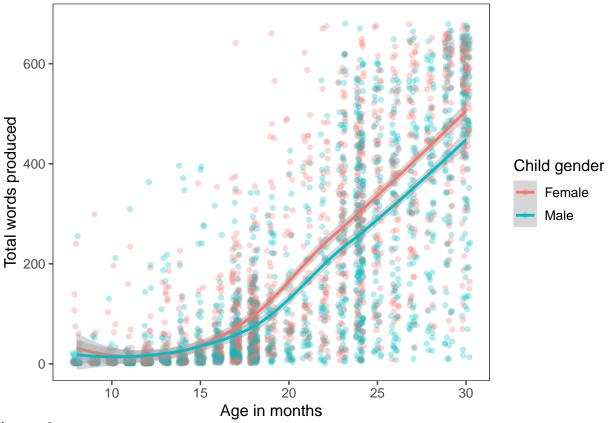


Figure 3

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.

measures, vocabulary scores were slightly negatively correlated with primary caregivers' education level, such that those parents without any college education reported higher vocabulary scores on both scales. A linear regression model with robust standard errors predicting comprehension scores with children's age and primary caregivers' education level (binned into categories of "High school diploma or less", "Some college education" and "College diploma or more³") as predictors shows main effects of both age ($\beta = 19.89$, p <

³ "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.

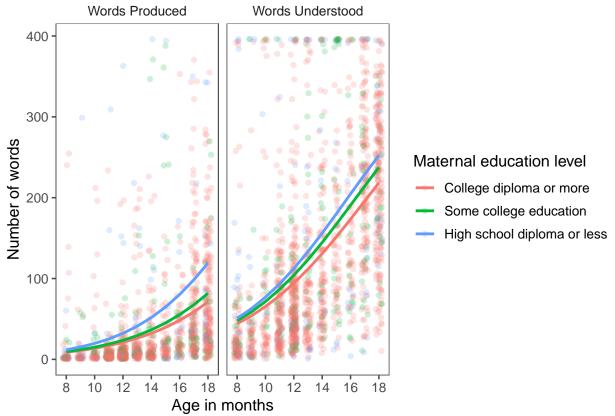


Figure 4

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.

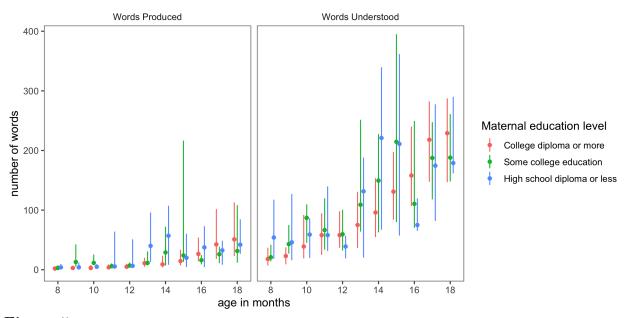


Figure 5 Median vocabulary production (left) and comprehension (right) scores by age on the WG form. Lines indicate span between first and third quartiles for each age.

0.001) and caregiver primary education ($\beta_{highschool} = 29.59$, p = 0.01). Similarly, a linear regression model with robust standard errors predicting production scores by children's age 375 and primary caregivers' education level shows main effects of age ($\beta = 7.82, p < 0.001$) and 376 caregiver primary education ($\beta_{highschool} = 28.86$, p = 0.002).

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The pattern of results seen in the WG sample is consistent with prior findings 378 indicating that respondents with lower levels of education attainment report higher 379 vocabulary comprehension and production on the CDI-WG form (Feldman et al., 2000; 380 Fenson et al., 1994). Although caregivers with lower levels of education attainment report 381 higher mean levels of vocabulary production and comprehension, median vocabulary scores 382 (which are more robust to outliers) show no clear pattern of difference across primary 383 caregiver education levels (Figure 5). This discrepancy between the regression effects and a 384 group-median analysis suggests that the regression effects described previously are driven in 385 part by differential interpretation of the survey items, such that a few lower-SES caregivers 386

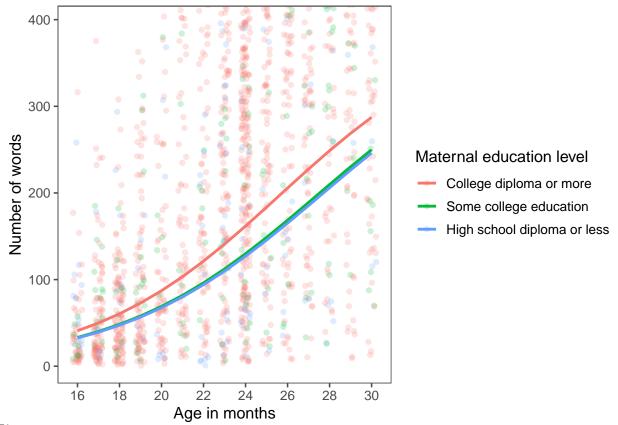


Figure 6

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.

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 390 al. (2000) and Frank et al. (2021) such that maternal education is positively associated with 391 children's reported vocabulary size (Figure 6). Because representation of caregivers without 392 a high school diploma is scarce (N = 18 out of a sample of 1,948), interpretation of the data 393 from this group is constrained. Nevertheless, as shown in Figure 6, a small but clear positive 394 association between maternal education and vocabulary score exists such that 395 college-educated caregivers report higher vocabulary scores than those of any other 396 education level. The implications from these data converge with previous findings which 397 indicate that parental education levels, often used as a metric of a family's socioeconomic 398 status, are related to children's vocabulary size through early childhood. 390

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

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Despite the large sample sizes we collected in the previous section, our current dataset from Web-CDI is, if anything, even more biased towards highly-educated White families than previous datasets. How can we recruit more diverse samples to remedy this issue? Here we discuss some potential routes forward. In this first effort we focus on collecting data from monolingual English-speaking families. While understanding that 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
Houwer, 2019), we focused here only on monolingual development, because collecting data
from multilingual populations introduces additional methodological considerations (e.g., how
to measure exposures in each language) that are not the focus of our work here.

417 Online data collection

Online recruitment methods, such as finding participants on platforms such as 418 Amazon Mechanical Turk, Facebook and Prolific, represent one possible route towards 419 assembling a large, diverse sample to take the Web-CDI. These methods allow researchers 420 depart from their typical geographical recruitment area much more easily than 421 paper-and-pencil administration. However, these recruitment methods are to our knowledge 422 largely untested with parent report measures of child language development. In a series of 423 data collection efforts, we used Web-CDI as a tool to explore these different channels of 424 recruitment. 425

In our first phase of data collection, we ran advertisements on Facebook which were 426 aimed at non-white families based on users' geographic locations (e.g., targeting cities which 427 have a higher than average representation of African Americans) or other profile features 428 (e.g., ethnic identification, interest in parenthood-related topics). Advertisements consisted of 429 an image of a child and a caption informing Facebook users of an opportunity to fill out a 430 survey on their child's language development and receive an Amazon gift card (Figure 7). 431 Upon clicking the advertisement, participants were redirected to a unique administration of 432 the Web-CDI, and they received \$5 upon completing the survey. This open-ended approach 433 to recruitment offered several advantages, namely that a wide variety of potential participants from specific demographic backgrounds can be reached on Facebook. However, 435 we also received many incomplete or otherwise unusable survey administrations, either from 436 Facebook users who clicked the link and decide not to participate, or those who completed 437



Figure 7

Example Facebook advertisement in Phase 1 of recent data collection.

the survey in an extremely short period of time (over half of all completed administrations,
Table 3).

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 demographic information about participants, allowing researchers to specify whom they would like to complete their studies. Prolific further has a built-in compensation infrastructure that handles monetary payments to participants, eliminating the need to

Table 3

Exclusions from recent data collection using Facebook and Prolific.

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%

disburse gift cards through Web-CDI.

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In the particular case of Web-CDI, the demographic information needed to determine
whether an individual was eligible to complete our survey (e.g., has a child in the correct age
range, lives in a monolingual household, etc.) was more specific than the information that
Prolific collects about their participant base. We therefore used a brief pre-screening
questionnaire to generate a list of participants who were eligible to participate, and
subsequently advertised the Web-CDI survey to those participants. Given that we were
interested only in reaching participants in the United 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 than that which we could obtain on Facebook.

Across both phases (Facebook and Prolific recruitment), we used the same exclusion

criteria as in the full Web-CDI sample to screen participants. A complete tally of all 459 excluded participants is shown in Table 3. In both the WG and WS surveys, exclusion rates 460 were high, amounting to 58% of participants who completed the survey. The high exclusions 461 rates were notably driven by an accumulation of survey administrations which participants 462 completed very quickly (in these analyses, defined as a completion taking less than 8.5 463 minutes). Many of the survey administrations excluded for fast completion had missing 464 demographic information reported: Among WG participants excluded for too-fast 465 completions, 93% did not report ethnicity, and among WS participants excluded for the same reason, 2% did not report ethnicity. After exclusions, full sample size was N = 115 WG467 completions and N = 126 completions. 468

The results from our recent data collection efforts focused on lower-SES, non-white 469 participants show overall similar patterns to the full Web-CDI sample in several regards. 470 Word production scores from both the WG and WS administrations reflect growing 471 productive vocabulary across the second and third years, with a very small gender effect 472 such that female children's vocabularies grow at a slightly faster rate than males' (Figure 8). 473 The relationship between caregivers' reported levels of maternal education and child's 474 vocabulary score is not as clear as it is in the full Web-CDI sample (Figure 9); however, 475 children of college-educated parents show slightly faster vocabulary growth than do children 476 of parents without any college degree. These patterns suggest that our data show similar 477 general patterns to other CDI datasets with other populations (Frank et al., 2021). 478

Importantly, recent data collection efforts showed a substantial improvement in 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 participants than the overall Web-CDI sample and the norms established by Fenson et al. (2007) (Figure 10). Black participants in particular showed a marked increase in representation, from 10.5% in the 2007 norms to 30.9% in the recent sample, while the

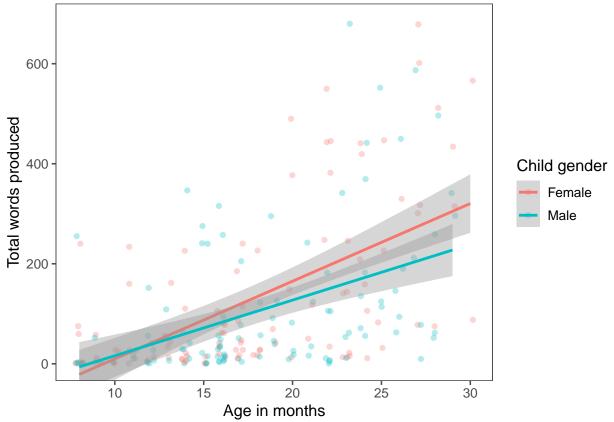


Figure 8

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.

proportion of white participants decreased from 73.3% in the 2007 norms to 50.2% in the recent sample. Representation on the basis of families' reported primary caregiver education also improved (Figure 10). Participants with only a high school diploma accounted for 31.5% of the recent sample as compared to 23.8% in the 2007 norms, and representation of those with a college diploma or more education decreased from 43.8% in the 2007 norms to 36.5% in the recent sample.

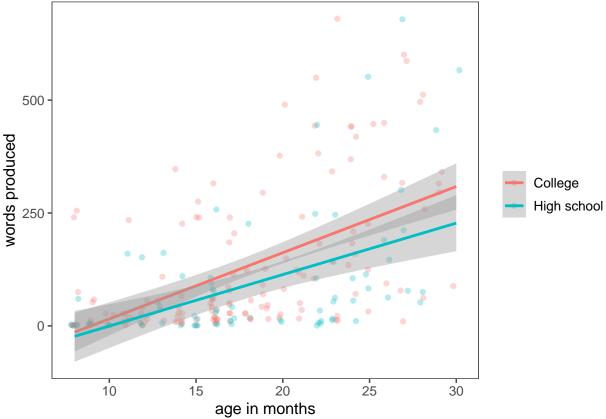


Figure 9

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.

91 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

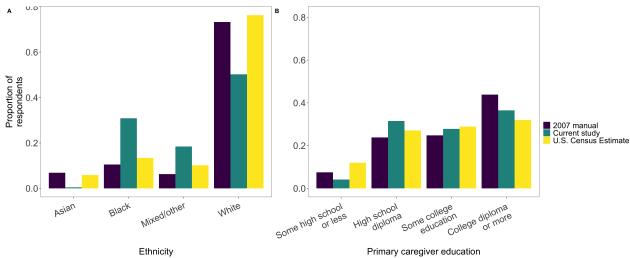


Figure 10

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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.

8.5 minutes, and of these quick completions, well over 90% were missing demographic information that is rarely missing in other administrations of the form. Determining the precise reasons for the high exclusion rate, and how (if at all) this (self-)selection may bias data reflecting demographic trends in vocabulary development, requires a more thorough assessment of who is submitting hastily-completed forms. This assessment is beyond the scope of the current study. However, all respondents who got to the end of the form were compensated regardless of how thoroughly they completed it, creating the possibility that some participants who clicked the anonymous link may not have been members of the population of interest, but rather were other individuals motivated by compensation.

Additionally, the exclusion rates described previously only provide information on those participants who did, at some point, submit a completed form, but many individuals clicked the advertisement link and did not subsequently continue on to complete the form. Without an in-depth exploration of who is clicking the link and why they might choose not

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.

In a similar vein, participants in this study were recruited through a targeted post on 517 social media, a technique that is considerably more anonymous than recruitment strategies 518 which entail face-to-face or extended contact between researchers and community members. 519 Online recruitment methods may not be suitable for all communities, especially when 520 researchers ask participants to report potentially sensitive information about the health and 521 development of their children (even when such information is stored anonymously). Our goal 522 here was to assess whether general trends in past literature could be recovered using such an 523 online strategy, but future research should take into account that other more personal 524 methods of recruitment, such as more direct community outreach or liaison contacts, may 525 improve participants' experiences and their willingness to engage with the study.

Finally, a significant limitation of the current data collection process is that many 527 people in the population of interest - particularly lower-income families - do not have reliable 528 internet access. Having participants complete the Web-CDI on a mobile device may alleviate 529 some of the issues caused by differential access to Wi-Fi, since the vast majority of American 530 adults own a smartphone (Pew Research Center, 2018 - NEED TO ADD MANUALLY TO REFS). Accordingly, improving Web-CDI's user experience on mobile platforms will be an important step towards ensuring that caregivers across the socioeconomic spectrum can 533 easily complete the survey. For smartphone users on pay-as-you-go plans, who may be 534 reluctant to use data to complete a study, a possible solution could be compensating 535 participants for the amount of "internet time" they incurred completing the form. 536

537 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, collect vocabulary development data using the MacArthur-Bates CDI. With traditional paper-based forms, combining insights from various research groups can prove challenging, as each group may have slightly different ways of formatting and managing data from CDI forms. By contrast, if all of these groups' data come to be stored in a single repository with a consistent database structure, data from disparate sources can easily be collated and analyzed in a uniform fashion. As such, a centralized repository such as Web-CDI provides a streamlined data-aggregation pipeline that facilitates cross-lab collaborations, multisite research projects and the curation of large datasets that provide more power to characterize the vast individual differences present in children's vocabulary development.

Beyond the goal of simply getting more data, we hope that Web-CDI can advance
efforts to expand the reach of vocabulary research past convenience samples into diverse
communities. A key question in the field of vocabulary development concerns the
mechanisms through which sociodemographic variables, such as race, ethnicity, income and
education are linked to group differences in vocabulary outcomes. Large,
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 data from 564 non-white, lower-SES communities in two phases. Several overall patterns emerged from the resulting data which we expected: vocabulary scores grew with age, providing a basic 566 validity check of the Web-CDI measure; females held a slight advantage in word learning over males; and children of parents with a college education showed slightly higher vocabulary 568 scores. Nonetheless, the insights from these data, while aligned with past norming studies, 569 are necessarily constrained by several features of our method. First, exclusion rates among 570 data collected on Facebook were very high, well over 50%, mostly due to a large quantity of 571 hasty completions. Second, a rigorous evaluation of the population-representativeness of 572 those who were counted in the final sample was not feasible here. 573

Web-based data collection can capture useful information about vocabulary
development from diverse communities, but future research will need to examine which
sampling methods can yield accurate, population-representative data that can advance our
understanding of the link between sociodemographic variation and variation in language
outcomes.

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