

1 **A systematic review of neuroimaging approaches to single-subject studies of language**  
2 **processing**

3 Aahana Bajracharya and Jonathan E. Peelle  
4 Department of Otolaryngology, Washington University in St. Louis, St. Louis, MO, USA  
5

6 **Corresponding Author:**

7 Aahana Bajracharya  
8 660 South Euclid Avenue, St. Louis, MO 63110  
9 Email address: [aahana@wustl.edu](mailto:aahana@wustl.edu)  
10

11 **Acknowledgments:** We thank Angela Hardi, Clinical Librarian at the Bernard Becker Medical  
12 Library, Washington University School of Medicine, for assisting with the database search and  
13 members of the Peelle lab for helping with the abstract screening process.  
14

15 **Conflict of Interest:** The authors report no conflict of interest.  
16

17 **Funding sources:** Grants R01 DC014281, R21 DC016086, R21 DC015884, and T32 EB014855  
18 from the US National Institutes of Health.  
19

20 **Keywords:** fMRI, language, reliability, single-subject, cognitive neuroscience  
21  
22

23 **Abstract**

24 Task-based functional magnetic resonance imaging (fMRI) has become the method of choice for  
25 studying localized function in the human brain. Functional MRI studies often rely on group-level  
26 results to derive conclusions about the neurobiology of language. However, doing so without  
27 accounting for the complexities of individual brains may reduce the validity of the findings.  
28 Furthermore, understanding brain organization in individuals is critically important for both  
29 basic science and clinical application. To assess the state of single-subject language localization  
30 in the functional neuroimaging literature, we carried out a systematic review of studies published  
31 through April 2020. Out of 977 papers identified through our search, 121 met our inclusion  
32 criteria for reporting single-subject fMRI results. Of these, 20 papers reported using a single-  
33 subject test-retest analysis to assess reliability. Specific metrics included overlap measures (like  
34 the Dice coefficient), correlation measures (like Intraclass Correlation Coefficient), Euclidean  
35 Distance between peak activation/center of mass, and Sensitivity/Specificity. These papers  
36 varied substantially in their experimental paradigms and stimuli, making more detailed  
37 comparisons impossible. In the absence of quantified reproducibility, results from paradigms  
38 used for single-subject language localization may need to be treated with caution. Incorporating  
39 reliability and validity measures in language mapping paradigms increases the likelihood that  
40 task-based activations are reproducible. Our search found that a relatively modest number of  
41 papers reporting single-subject results quantified single-subject reliability. Future endeavors to  
42 optimize the localization of language networks in individuals will benefit from the broader  
43 reporting of reliability metrics for different tasks and acquisition parameters.

44 **Introduction**

45 Historically, much of our understanding of the neurobiology of language has come from lesion  
46 studies and the differing profiles of patients with acquired aphasia. Subsequent advances in  
47 functional neuroimaging methods have, helpfully, broadened our view of brain regions involved  
48 in language processing. Many of these endeavors rely on the conclusions drawn on a group level,  
49 even though language networks might vary from person to person. Thus, group results may  
50 suggest an organization that does not accurately represent any individual's language map  
51 (Fedorenko et al. 2010; Mahowald & Fedorenko 2016). Individual and group-level analyses are  
52 therefore complementary approaches that inform different aspects of how we understand  
53 language capabilities.

54 Characterizing language processing in individuals is not just a theoretical concern:  
55 Numerous clinical studies have shown evidence for atypical language organization due to  
56 various conditions, including epilepsy (Baciu et al. 2003; Gould et al. 2016; Lee et al. 2008),  
57 aphasia (Khateb et al. 2004), vascular malformations (Hakyemez et al. 2006; Pouratian et al.  
58 2002), brain injury and long-standing tumors (Avramescu-Murphy et al. 2017; Kośła et al. 2015;  
59 Partovi et al. 2012; Ruff et al. 2008), and sensory deficits caused by congenital blindness (Röder  
60 et al. 2002; Roland et al. 2013). Non-clinical conditions such as left-handedness have been  
61 associated with an increased incidence of atypical language dominance (Acioly et al. 2014).  
62 Additionally, language processing differs for primary and secondary languages (Dehaene et al.  
63 1997; Polczynska et al. 2016; Polczynska et al. 2017; Tomasino et al. 2014). Multilingual  
64 individuals also often demonstrate the recruitment of additional brain areas for language  
65 switching (Sierpowska et al. 2013; Tomasino et al. 2014). Given the complex nature of language  
66 representation in the brain, failure to map language regions effectively in preoperative mapping  
67 could mean a tradeoff between losing language function altogether versus capability in only one  
68 language.

69 A related and long-standing concern in cognitive neuroscience has been the degree to  
70 which fMRI-based activations, generally, are reliable (Bennett & Miller 2010; Elliott et al. 2020;  
71 Gorgolewski et al. 2013; McGonigle 2012; McGonigle et al. 2000; Noble et al. 2019; Smith et al.  
72 2005). As with many areas of science, there is an increasing interest in measuring and improving  
73 reliability and reproducibility of neuroimaging research (Botvinik-Nezer et al. 2020; Button et al.  
74 2013; Nosek & Lakens 2014; Poldrack et al. 2017; Simmons et al. 2011). Relatedly, there has  
75 been a recent re-emergence of interest in individual variability in brain organization (Gordon et  
76 al. 2017; Laumann et al. 2015; Poldrack et al. 2015)— which assumes (implicitly, if not  
77 explicitly) that differences in brain maps across individuals reflect true neural differences and not  
78 measurement error. Thus, accurate single-subject language mapping is essential in two contexts.  
79 First, it tells us about individual brain organization; and second, because measurement accuracy  
80 in single subjects affects the accuracy of group-level analyses. However, the degree to which  
81 neuroimaging studies of language localization have assessed the reliability of experimental  
82 paradigms is unclear.

83 To characterize the current state of the field with respect to single-subject language  
84 localization, we performed a systematic review of fMRI studies reporting single-subject results.  
85 Our goals were to document approaches used for assessing the reliability of single-subject  
86 results, place language studies in a broader context of fMRI reliability, and, if possible, identify

87 potential design choices associated with improved single-subject reliability.

## 88 **Materials and Methods**

89 The information gathered in this systematic review was structured by following the PRISMA  
90 statement (Liberati et al. 2009), summarized in **Figure 1**. Supplemental materials, including data  
91 and analysis scripts, are available from <https://osf.io/x692b/>. We used SunburstR package in R to  
92 create the figures (Bostock et al. 2020; Team 2013).

## 93 **Search Methods Statement**

94 We searched published literature using strategies (Search Strategy in Supplemental materials)  
95 designed by a medical librarian for the concepts of functional magnetic resonance imaging  
96 (fMRI), speech or language mapping, and brain mapping. These strategies were established  
97 using a combination of controlled vocabulary terms and keywords and were executed in Ovid-  
98 Medline, Embase, Scopus, Cochrane Register of Controlled Trials (CENTRAL), and  
99 Clinicaltrials.gov. We verified the effectiveness of the search strategies based on how well a set  
100 of predefined benchmark papers were captured by the search. All searches were performed on  
101 April 8<sup>th</sup>, 2020. Results were exported to EndNote, and duplicate citations were removed, leaving  
102 970 unique citations for analysis. Database-supplied limits for English were used.

## 103 **Additional literature**

104 The initial search did not capture seven previously identified benchmark papers due to missing  
105 keywords in the title or abstract. These were added to the final list of papers, which resulted in a  
106 total of 977 papers.

## 107 **Eligibility Criteria**

108 We selected all published papers before April 8<sup>th</sup>, 2020, that met the following criteria:

- 109 • Used fMRI as an independent modality or in conjunction with other modalities
- 110 • Primary research literature in adults
- 111 • Language task used in the experimental design
- 112 • Performed single-subject level analysis
- 113 • Reported task-based single-subject maps or quantification of single-subject results

114 Specific search terms are available in supplemental materials. The resulting papers from the  
115 database search then underwent abstract screening and full-text screening using the procedure  
116 discussed below.

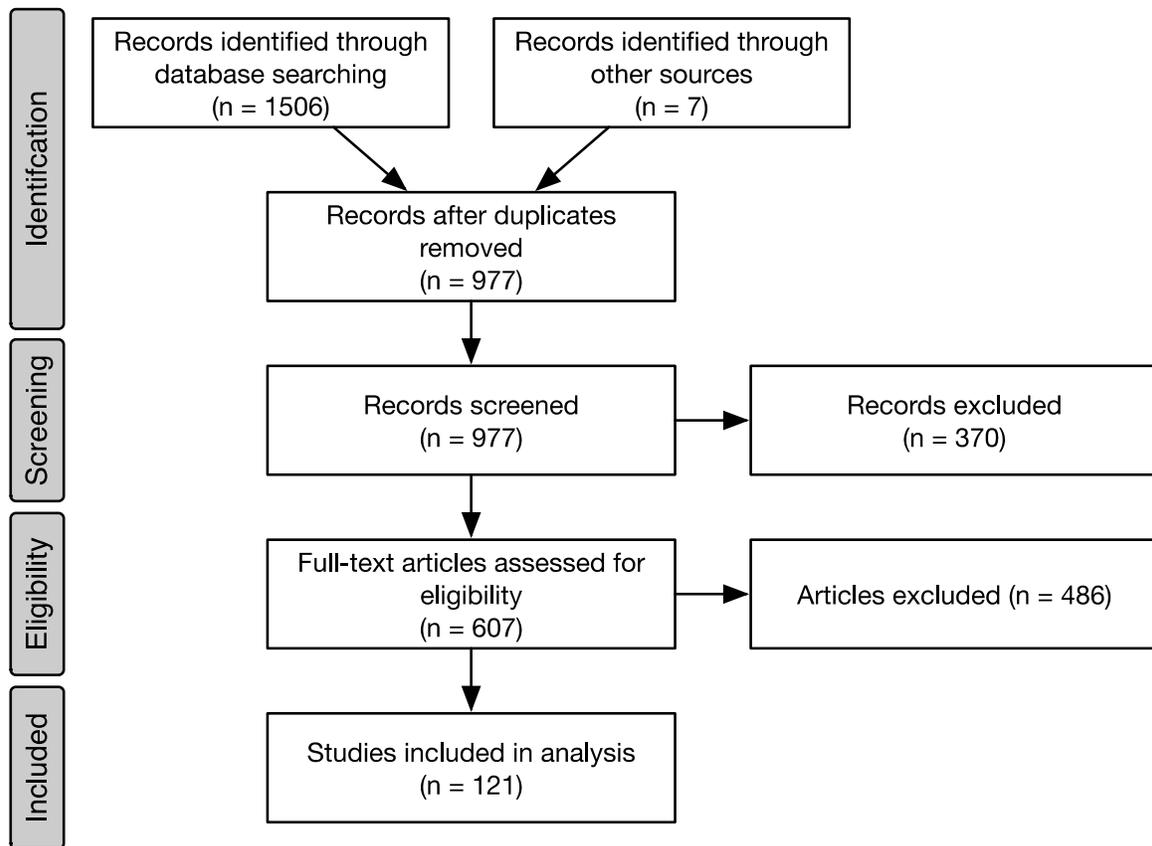
## 117 **Abstract Screening**

118 We screened abstracts obtained after the literature search to exclude conference abstracts,  
119 reviews, technical notes, clinical trials, and other non-primary literature. Articles that were  
120 unavailable after a Google Scholar search, duplicate entries, incorrect citations, empty results for  
121 abstract contents, studies conducted on children, and abstracts that did not mention fMRI use  
122 were also excluded. The extent to which fMRI was used as an imaging modality was not always  
123 clear from the abstract alone. Therefore, papers with abstracts that mentioned task-based fMRI,

124 or cited fMRI results, passed the abstract screening step. A group of individuals with an  
125 academic background in neuroscience assisted with the abstract level screening. After an initial  
126 categorization, abstract eligibility was verified by the first author. All the abstracts were  
127 reviewed by at least one person and checked by the first author.

### 128 **Full-text screening**

129 We then screened the contents of the articles that passed abstract screening. Papers were  
130 excluded from the final list if they did not contain quantified single subject task-based fMRI  
131 results and only reported group-level results. The finalized papers were screened and categorized  
132 based on imaging modalities, reliability metrics, language tasks used, and clinical condition of  
133 research participants. To the extent possible, we categorized the tasks using labels from the  
134 Cognitive Atlas (Poldrack et al. 2011).



135

136

**Figure 1:** PRISMA diagram summarizing the literature search.

### 137 **Reliability measures**

138 A common way of quantifying a neuroimaging study's reliability is to assess the test-retest  
139 reproducibility. Assuming brain networks have remained stable, performing the same task should  
140 result in a similar pattern of brain activity, with differences attributable to measurement error. A  
141 concern with these metrics is the amount of data available to carry out analyses or the technical  
142 considerations of repeating an experiment. Nevertheless, including measures for reproducibility

143 may increase confidence in the findings or establish precedence towards good research practice.  
 144 The following are the most common measures used by the papers included in this review.

- 145 • **Lateralization Index (LI)** is used as a comparative measure of language-related  
 146 activations between the hemispheres of the brain. Although not a reliability measure on  
 147 its own, a considerable number of papers rely on the reproducibility of LI as a reliability  
 148 metric (Agarwal et al. 2018; Benjamin et al. 2017; Fernandez et al. 2003; Knecht et al.  
 149 2003; Nettekoven et al. 2018; Otzenberger et al. 2005; Voyvodic 2012; Wilson et al.  
 150 2018). A score of 1 indicates fully left-lateralized activation, a score of -1 indicates fully  
 151 right-lateralized activation, and a score of 0 refers to bilateral (i.e., non-lateralized)  
 152 activation.

153 LI is a metric of interest in reliability because of the typical dominance of  
 154 language in the left hemisphere. Lateralization of function may differ in relation to the  
 155 complexity of the stimulus used in the task design (Peelle 2012). A task involving  
 156 hearing tones might only activate the bilateral auditory cortex whereas stimuli with more  
 157 complex language requirements might have activations localized to the left hemisphere.  
 158 Hence, LI can be an effective way to communicate stimulus-based differences in brain  
 159 activations.

160 Studies using the Wada procedure as the primary language mapping tool often  
 161 rely on the reproducibility of LI as their only reliability measure. However, this will be of  
 162 limited use since the information conveyed by LI is not enough to fully direct surgical  
 163 decisions as it ignores localization. Moreover, the robustness and strength of different  
 164 language tasks might also contribute to the variability (Bradshaw et al. 2017b). Thus, LI  
 165 is best used in conjunction with other metrics.  
 166

$$167 \quad LI = \frac{\sum \text{left activations} - \sum \text{right activations}}{\sum \text{left activations} + \sum \text{right activations}}$$

- 168 • **Overlap measures** such as Dice Coefficient measures the overlap of the number of  
 169 active voxels across scan sessions separated by time. The Dice coefficient for any two  
 170 sessions, as calculated by the following equation, ranges from 0 to 1, with 0 indicating no  
 171 overlap and 1 indicating complete overlap (Crum et al. 2006). Several studies included in  
 172 this review have implemented the Dice (or related overlap measure). An example of a  
 173 related measure is the Reproducibility Index, as mentioned in Maldjian et al. (2002).  
 174 Here, the metric is obtained by calculating the pairwise ratio of the probability-weighted  
 175 intersection volume divided by the union volume of surviving activation clusters. Overlap  
 176 metrics such as Dice are some of the most intuitive methods of accounting for  
 177 reproducibility. However, factors such as absolute and relative voxel-wise thresholds,  
 178 cluster sizes, and focus on a priori language regions could affect this measure (Wilson et  
 179 al. 2017).

$$180 \quad \text{Dice coefficient} = 2 \frac{\text{Number of overlapping voxels}}{\text{Voxels in first session} + \text{Voxels in second session}}$$

- 181 • **Intra-class Correlation Coefficient (ICC)** is calculated by dividing the difference  
 182 between and within-subjects mean sum of squares by their sum (Fernandez et al. 2003).  
 183 ICC ranges from -1 to 1, with ICC < 0 indicating no agreement and ICC of 1 indicating

184 perfect agreement (Nettekoven et al. 2018). It can be used to measure test-retest  
 185 reliability at a voxel or an ROI for a chosen level of activation. This metric provides a  
 186 measure of the contribution of individual-level differences in a group result. However, it  
 187 is vital to treat this measure with caution since it combines information from between-  
 188 subject and between-sessions variances (i.e., the same ICC value can result from different  
 189 activation patterns resulting from inadequate models) (Gorgolewski et al. 2013).

$$191 \quad ICC = \frac{\textit{Between subject variance} - \textit{Within subject variance}}{\textit{Between subject variance} + (k - 1) \textit{Within subject variance}},$$

192 where  $k$  is the number of test sessions.

- 193 • **Euclidean Distance (ED)** is commonly used to quantify the distance between peak  
 194 activation across sessions or different task types (Agarwal et al. 2018; Nettekoven et al.  
 195 2018; Voyvodic 2012). Localization accuracy can be determined based on how close the  
 196 activation peaks are for subsequent sessions. This metric can be highly susceptible to  
 197 shifts in activation patterns due to subject motion across sessions.

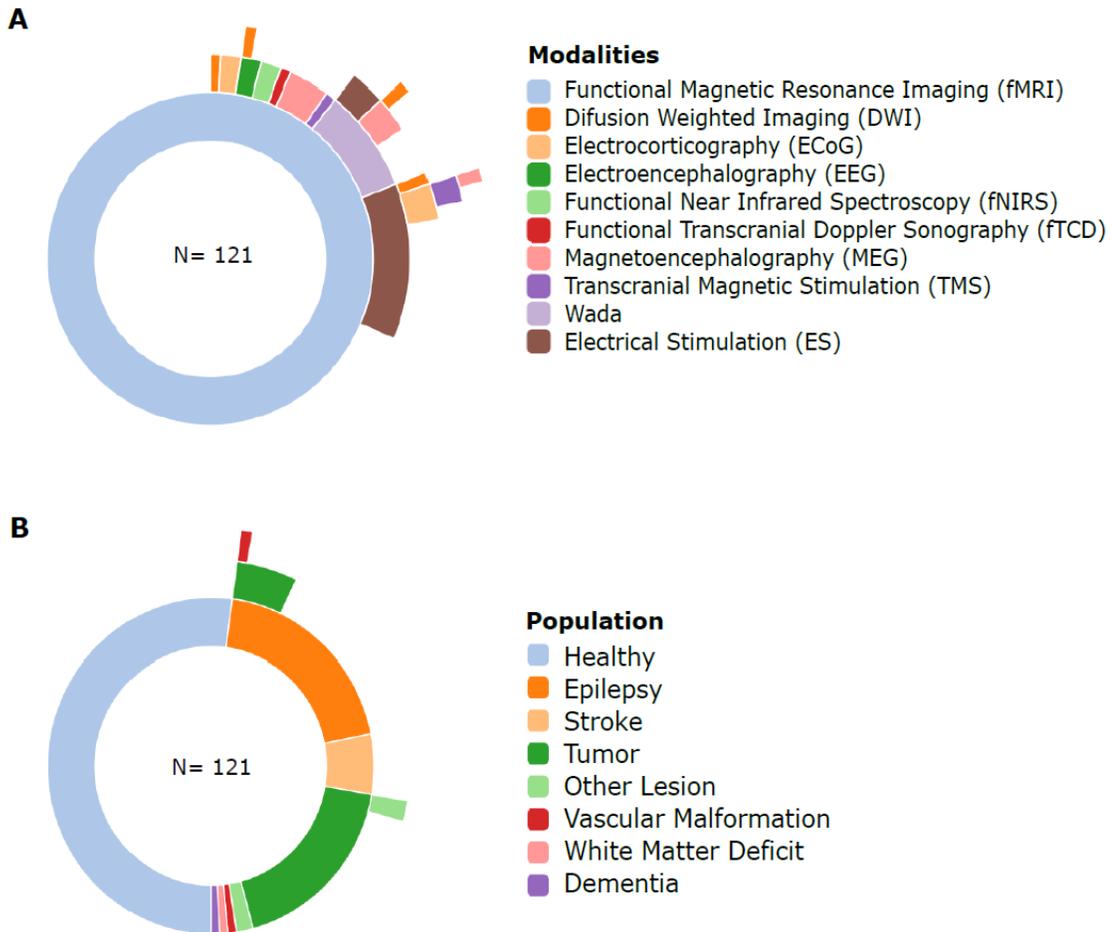
$$199 \quad ED = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2 + (z_1 - z_2)^2},$$

200 where,  $x_{1,2}, y_{1,2}, z_{1,2}$  represent coordinates in 3D space.

## 202 Results

203 Our literature search resulted in 977 unique papers, of which 121 met our inclusion criteria of  
 204 using single-subject level fMRI to study language processing. Thirty-eight out of the 121 studies  
 205 used more than one modality to carry out language mapping. The distribution of the modalities  
 206 used by the papers is presented in **Figure 2A**. Language localization studies in a clinical setting  
 207 are sometimes conducted in the context of pre-surgical planning. Most of the clinical studies that  
 208 met our inclusion criteria used fMRI as a preliminary method for language localization but relied  
 209 on the results of invasive mapping methods (such as electrical stimulation) to carry out surgical  
 210 planning. Additionally, the feasibility of using fMRI in conjunction with electrical stimulation to  
 211 predict language dominance has also been tested in post-surgical cases (Peck et al. 2009).  
 212 Several publications in our literature search have also compared various modalities under the  
 213 same experimental paradigm to demonstrate the relative effectiveness of the individual  
 214 approaches. Studies with Electrocorticography (ECoG), fMRI, and electrical stimulation reported  
 215 that ECoG and fMRI results were better than electrical stimulation (Genetti et al. 2015; Tie et al.  
 216 2009). Similarly, since electrical stimulation is the current gold-standard in brain mapping,  
 217 studies have used it as a metric to optimize and standardize fMRI protocols (Rutten et al. 2002;  
 218 Wilson et al. 2018). Moreover, numerous studies have used multimodal approaches to  
 219 demonstrate the benefits of combining the strengths of the individual modalities such as fMRI  
 220 and Magnetoencephalography (MEG), Transcranial Magnetic Stimulation (TMS) and fMRI  
 221 (Kononen et al. 2015), simultaneous Electroencephalography (EEG) and ultra-high field MRI  
 222 (Grouiller et al. 2016) and, fMRI and Diffusion Weighted Imaging (DWI) (Tomasino et al.  
 223 2014).

224 The distribution of the clinical conditions represented in the papers is summarized in  
 225 **Figure 2B**. Out of the 121 studies reporting single-subject fMRI results, 58 included clinical  
 226 populations, most of which were multimodal studies. These studies included a wide variety of  
 227 clinical conditions, epilepsy and tumor being the most common. The study population's  
 228 distribution highlights the importance of reliable single-subject language mapping methods for  
 229 both clinical and basic research.



230  
 231 **Figure 2:** Each circle or part of a circle in the Sunburst plot indicates the proportion of studies that belong  
 232 to the subgroup represented in the legend. **A.** The full inner circle shows the number of studies that used  
 233 fMRI as one of the imaging modalities. Each subsequent layer represents the proportions of studies with  
 234 each added modality. **B.** The proportions represent the distribution of population subtypes in the papers.  
 235 Subsequent layers indicate studies that investigated individuals from more than one population. Details on  
 236 the number of papers for each category is presented in the supplemental information.

237 As noted above, we were particularly interested in how many papers reported reliability  
 238 measures. Of the 121 papers reporting single subject results, 20 reported test-retest sessions and  
 239 discussed reliability measures as summarized in **Table 1**. The duration between test and retest  
 240 sessions in these papers ranged from as close as a few minutes (consecutive test and retest on the  
 241 same day) to a few years. Following Wilson et.al (2017), we make a distinction between validity  
 242 (effectiveness of tasks to activate known language regions) and reliability (test-retest

243 reproducibility of data). Most common validity measures were hemispheric lateralization  
 244 (16/20), volume of activation (4/20), and sensitivity/specificity calculations (3/20). Meanwhile,  
 245 most common reliability measures were overlap metrics (e.g., Dice) (16/20), correlation  
 246 measures (e.g., Intraclass Correlation Coefficient) (7/20), and comparison of the distance  
 247 between peak activations (5/20).

248

Paper	Lateralization variability	Activation volume	Sensitivity/ Specificity	Overlap metric	Correlation metric	Activation Distance
Binder et.al (1995)	X					
Maldjian et.al (2002)	X			X		
Rutten et.al (2002)	X	X		X		
Fernández et.al (2003)	X			X	X	
Knecht et.al (2003)	X			X		
Otzenberger et.al (2005)	X			X		
Harrington et.al (2006)	X			X		
Jansen et.al (2006)	X			X	X	
Chen et.al (2007)			X			
Rau et.al (2007)				X		
Voyvodic (2012)	X			X		X
Gorgolewski et.al (2013)				X	X	X
Mahowald & Fedorenko (2016)	X	X			X	
Benjamin et.al (2017)	X			X		
Wilson et.al (2017)	X		X	X		
Agarwal et.al (2018)	X					X
Nettekoven et.al (2018)	X			X	X	X
Wilson et.al (2018)	X	X		X	X	X
Paek et.al (2019)				X	X	
Yen et.al (2019)	X	X	X	X		

249 **Table 1:** Summary of validity and reliability measures used in the papers that reported test-retest  
 250 sessions.

251 A quantitative comparison of the studies is not possible due to the variability in tasks,  
 252 study population, duration between scans, and thresholding measures (detailed in  
 253 **Supplementary Table 1**). There was a wide range of expressive and receptive tasks used in a  
 254 multi-task setting (Acioly et al. 2014; Arora et al. 2009; Seghier et al. 2004; Tailby et al. 2017).  
 255 The most common tasks among these studies were verbal fluency tasks (such as naming and  
 256 word generation), while less common tasks involved connected speech. The 20 papers that  
 257 discussed test-retest reliability used block design for their experiments but differed in the  
 258 technical execution.

259 Overlap measures were used in 16 of the 20 papers, with 8 using Dice. The 8 Dice papers  
 260 reported values ranging from 0.34–0.66.

261 **Discussion**

262 Accurate measurements of regional brain activation in individual participants are essential for  
263 clinical research and basic science. In the context of group studies, individual variability in task-  
264 based responses has been noted in several domains (Van Horn et al. 2008). The focus of our  
265 current review was on studies reporting fMRI-based language localization in single subjects.  
266 Given the diverse range of language research that spans the clinical and basic science domains,  
267 the search results that we have presented might not have captured all the relevant literature.  
268 However, the need for reproducible and robust results is essential in any research outcome. The  
269 metrics discussed in this paper can also serve as a primer for the most common ways in which  
270 reliability metrics can be used while reporting neuroimaging results, and we hope to encourage  
271 wider adoption of such analyses.

272         Although test-retest reproducibility is not the only measure of accuracy, it is widely used  
273 and directly addresses within-subject replicability. Of the papers identified in our search,  
274 approximately 1/6 included some measure of test-retest reliability. However, this represents a  
275 modest number of studies (20) that vary considerably in the specific language paradigm,  
276 participant population, and amount of data collected. Moreover, the choice of metric used to  
277 establish reliability was not the same across these studies. Thus, it may prove challenging to  
278 generalize existing findings of test-retest reliability to new paradigms or populations. Obtaining  
279 an agreement on a standardized approach of quantifying reliability in neuroimaging results  
280 would enhance the credibility of research findings. Given that most conclusions are drawn from  
281 thresholded statistical maps, the Dice overlap measure is an appealing candidate.

282         Across a variety of paradigms, we found that average Dice coefficients ranged from  
283 0.34–0.66. This range is roughly comparable to that reported by Bennett and Miller (2010), who  
284 across a large number of tasks (most not language tasks), report a range of average Dice  
285 coefficients from 0.23–0.79. As Bennett and Miller highlight, many factors contribute to  
286 reliability of fMRI studies, including differences in acquisition, analysis, paradigm, and  
287 participants. The number of potential permutations among these factors typically make  
288 controlled comparisons impossible. That is, although an individual study may ask “among these  
289 four paradigms, which provides the strongest reliability?”, it is far more difficult to answer,  
290 “among all the tasks, analysis pipelines, and acquisition parameters available, which provides the  
291 strongest reliability?” (which we would all like to know!). However, the variability in reliability  
292 measures (such as Dice) suggest that some approaches are more reliable than others. As  
293 suggested above, one approach would be to more widely adopt reporting of reliability measures  
294 to facilitate optimizing these protocols. More simply, labs might internally use such metrics  
295 during task development, to steer them away from tasks that have poor reliability.

296         In the context of modest test-retest reliability, it is important to note that the issue of  
297 single-subject reliability also extends to group-level studies. If the outcome of interest is a group-  
298 level univariate map with a relatively large number of participants, inaccuracies in individual  
299 participants may have little effect on the result. However, researchers interested in explaining  
300 individual differences in brain activation patterns—for example, due to age, language status,  
301 hearing loss, etc.—rely on the accuracy of both neural and non-neural estimates of data at an  
302 individual level. At a minimum, inaccuracies in measuring brain activity in individual  
303 participants will hurt the ability of researchers to detect these effects of interest, and more  
304 worryingly, they may lead to spurious findings. A related concern applies to multivariate

305 analyses. Even when these analyses are not explicitly designed to localize activity in individual  
306 participants, multivariate tests are typically conducted in single subjects. Error in measuring  
307 responses that account for individual differences will likely decrease the accuracy of these  
308 analyses.

309         The high spatial resolution and noninvasiveness of fMRI enable it to complement other  
310 language localization approaches to target potentially more subtle or complex functions than  
311 standard clinical practices (Austermuehle et al. 2017; Baciú et al. 2003; Bizzi et al. 2008).  
312 However, using fMRI as an independent tool for localizing language for clinical purposes is still  
313 in its early stages. Patient and methodological challenges need to be addressed to do so  
314 (Beisteiner et al. 2019; Bradshaw et al. 2017a; Bradshaw et al. 2017b; Seghier 2008). An  
315 improvement in the reliability of language localization with fMRI might not replace invasive  
316 procedures entirely but can help identify subjects that can be assisted without invasive  
317 procedures. For those that do undergo invasive procedures, fMRI-based language localization  
318 can also assist in monitoring post-surgical recovery.

319         An important point is that all the studies identified by our search that discussed test-retest  
320 reliability carried out univariate analyses. Multivariate analyses are increasingly used to study  
321 individual differences (Woo et al. 2017), and may well provide reliability that exceeds traditional  
322 univariate approaches (Kragel et al. 2021). With respect to localization, multivariate approaches  
323 can vary in their spatial specificity, but searchlight approaches (Etzel et al. 2013; Kriegeskorte et  
324 al. 2006) provide an approach for conducting multivariate analyses throughout the brain.  
325 Multivariate approaches may therefore prove to be a valuable approach for improving reliability  
326 of language localization.

327         Finally, another avenue that facilitates assessing reliability is to make data freely  
328 available. Sharing original data sets would enable researchers to conduct their own measures of  
329 reliability across studies. Increasing awareness of data sharing benefits is occurring across  
330 scientific disciplines (Poldrack & Gorgolewski 2014), and publicly available infrastructure for  
331 sharing neuroimaging datasets continues to improve (Poldrack et al. 2013). As illustrated in our  
332 findings, a wide variety of tasks, populations, and metrics currently exist, making qualitative and  
333 quantitative comparisons across studies challenging.

334         In conclusion, we found that a relatively small number of papers investigating the  
335 neurobiology of language—all of which used univariate analysis methods—have assessed the  
336 test-retest reliability of single-subject fMRI paradigms. Increased attention to this issue can  
337 improve the accuracy and replicability of findings in multiple domains. Some concrete steps  
338 towards addressing these concerns could be making reliability metrics such as the Dice  
339 coefficient a standard part of analyses, reporting both single-subject and group level results to  
340 allow transparency, and making data freely available so that researchers can reproduce results or  
341 conduct their own reliability analyses.

## 342 **Supplementary Materials**

- 343         1. Database search strategies
- 344         2. Database search results
- 345         3. Table with detailed summary of metrics
- 346         4. PRISMA checklist
- 347         5. Code to generate figures

348 **References**

- 349 Acioly, M. A., A. Gharabaghi, C. Zimmermann, M. Erb, S. Heckl & M. Tatagiba. 2014.  
 350 Dissociated language functions: a matter of atypical language lateralization or cerebral  
 351 plasticity? *J Neurol Surg A Cent Eur Neurosurg* 75.64-9.
- 352 Agarwal, S., J. Hua, H. I. Sair, S. Gujar, C. Bettegowda, H. Lu & J. J. Pillai. 2018. Repeatability  
 353 of language fMRI lateralization and localization metrics in brain tumor patients. *Hum*  
 354 *Brain Mapp* 39.4733-42.
- 355 Arora, J., K. Pugh, M. Westerveld, S. Spencer, D. D. Spencer & R. Todd Constable. 2009.  
 356 Language lateralization in epilepsy patients: fMRI validated with the Wada procedure.  
 357 *Epilepsia* 50.2225-41.
- 358 Austermuehle, A., J. Cocjin, R. Reynolds, S. Agrawal, L. Sepeta, W. D. Gaillard, K. A.  
 359 Zaghoul, S. Inati & W. H. Theodore. 2017. Language functional MRI and direct cortical  
 360 stimulation in epilepsy preoperative planning. *Ann Neurol* 81.526-37.
- 361 Avramescu-Murphy, M, E Hattingen, M-T Forster, A Oszvald, S Anti, S Frisch, MO Russ & A  
 362 Jurcoane. 2017. Post-surgical language reorganization occurs in tumors of the dominant  
 363 and non-dominant hemisphere. *Clinical Neuroradiology* 27.299-309.
- 364 Baciú, M. V., J. M. Watson, K. B. McDermott, R. D. Wetzel, H. Attarian, C. J. Moran & J. G.  
 365 Ojemann. 2003. Functional MRI reveals an interhemispheric dissociation of frontal and  
 366 temporal language regions in a patient with focal epilepsy. *Epilepsy & Behavior* 4.776-  
 367 80.
- 368 Beisteiner, Roland, Cyril Pernet & Christoph Stippich. 2019. Can we standardize clinical  
 369 functional neuroimaging procedures? *Frontiers in Neurology* 9.1153.
- 370 Benjamin, C. F., P. D. Walshaw, K. Hale, W. D. Gaillard, L. C. Baxter, M. M. Berl, M.  
 371 Polczynska, S. Noble, R. Alkawadri, L. J. Hirsch, R. T. Constable & S. Y. Bookheimer.  
 372 2017. Presurgical language fMRI: Mapping of six critical regions. *Hum Brain Mapp*  
 373 38.4239-55.
- 374 Bennett, Craig M & Michael B Miller. 2010. How reliable are the results from functional  
 375 magnetic resonance imaging? *Annals of the New York Academy of Sciences* 1191.133-  
 376 55.
- 377 Binder, J. R. Rao S. M. Hammeke T. A. Frost J. A. Bandettini P. A. Jesmanowicz A. Hyde J. S.  
 378 1995. Lateralized human brain language systems demonstrated by task subtraction  
 379 functional magnetic resonance imaging. *Archives of Neurology* 52.593-601.
- 380 Bizzi, Alberto, Valeria Blasi, Andrea Falini, Paolo Ferroli, Marcello Cadioli, Ugo Danesi,  
 381 Domenico Aquino, Carlo Marras, Dario Caldiroli & Giovanni Broggi. 2008. Presurgical  
 382 functional MR imaging of language and motor functions: validation with intraoperative  
 383 electrocortical mapping. *Radiology* 248.579-89.
- 384 Bostock, Mike , Kerry Rodden, Kevin Warne, Kent Russell, Florian Breitwieser & CJ Yetman.  
 385 2020. sunburstR. CRAN.
- 386 Botvinik-Nezer, Rotem, Felix Holzmeister, Colin F Camerer, Anna Dreber, Juergen Huber,  
 387 Magnus Johannesson, Michael Kirchler, Roni Iwanir, Jeanette A Mumford & R Alison  
 388 Adcock. 2020. Variability in the analysis of a single neuroimaging dataset by many  
 389 teams. *Nature*.1-7.
- 390 Bradshaw, Abigail R, Dorothy VM Bishop & Zoe VJ Woodhead. 2017a. Methodological  
 391 considerations in assessment of language lateralisation with fMRI: a systematic review.  
 392 *PeerJ* 5.e3557.

- 393 Bradshaw, Abigail R, Paul A Thompson, Alexander C Wilson, Dorothy VM Bishop & Zoe VJ  
394 Woodhead. 2017b. Measuring language lateralisation with different language tasks: a  
395 systematic review. *PeerJ* 5.e3929.
- 396 Button, K. S., J. P. Ioannidis, C. Mokrysz, B. A. Nosek, J. Flint, E. S. Robinson & M. R.  
397 Munafo. 2013. Power failure: why small sample size undermines the reliability of  
398 neuroscience. *Nature Reviews Neuroscience* 14.365-76.
- 399 Crum, William R, Oscar Camara & Derek LG Hill. 2006. Generalized overlap measures for  
400 evaluation and validation in medical image analysis. *IEEE transactions on medical*  
401 *imaging* 25.1451-61.
- 402 Dehaene, Stanislas, Emmanuel Dupoux, Jacques Mehler, Laurent Cohen, Eraldo Paulesu,  
403 Daniela Perani, Pierre-Francois Van de Moortele, Stéphane Lehericy & Denis Le Bihan.  
404 1997. Anatomical variability in the cortical representation of first and second language.  
405 *NeuroReport* 8.3809-15.
- 406 Elliott, M. L., A. R. Knodt, D. Ireland, M. L. Morris, R. Poulton, S. Ramrakha, M. L. Sison, T.  
407 E. Moffitt, A. Caspi & A. R. Hariri. 2020. What Is the Test-Retest Reliability of Common  
408 Task-Functional MRI Measures? New Empirical Evidence and a Meta-Analysis. *Psychol*  
409 *Sci* 31.792-806.
- 410 Etzel, J. A., J. M. Zacks & T. S. Braver. 2013. Searchlight analysis: promise, pitfalls, and  
411 potential. *NeuroImage* 78.261-9.
- 412 Fedorenko, Evelina, Po-Jang Hsieh, Alfonso Nieto-Castañón, Susan Whitfield-Gabrieli & Nancy  
413 Kansiwhser. 2010. New method for fMRI investigations of language: Defining ROIs  
414 functionally in individual subjects. *Journal of Neurophysiology* 104.1177-94.
- 415 Fernandez, G, K Specht, S Weis, I Tendolkar, M Reuber, J Fell, P Klaver, J Ruhlmann, J Reul &  
416 CE Elger. 2003. Intrasubject reproducibility of presurgical language lateralization and  
417 mapping using fMRI. *Neurology* 60.969-75.
- 418 Genetti, M., R. Tyrand, F. Grouiller, A. M. Lascano, S. Vulliemoz, L. Spinelli, M. Seeck, K.  
419 Schaller & C. M. Michel. 2015. Comparison of high gamma electrocorticography and  
420 fMRI with electrocortical stimulation for localization of somatosensory and language  
421 cortex. *Clin Neurophysiol* 126.121-30.
- 422 Gordon, E. M., T. O. Laumann, A. W. Gilmore, D. J. Newbold, D. J. Greene, J. J. Berg, M.  
423 Ortega, C. Hoyt-Drazen, C. Gratton, H. Sun, J. M. Hampton, R. S. Coalson, A. L.  
424 Nguyen, K. B. McDermott, J. S. Shimony, A. Z. Snyder, B. L. Schlaggar, S. E. Petersen,  
425 S. M. Nelson & N. U. F. Dosenbach. 2017. Precision Functional Mapping of Individual  
426 Human Brains. *Neuron* 95.791-807 e7.
- 427 Gorgolewski, K. J., A. J. Storkey, M. E. Bastin, I. Whittle & C. Pernet. 2013. Single subject  
428 fMRI test-retest reliability metrics and confounding factors. *NeuroImage* 69.231-43.
- 429 Gould, L., M. J. Mickleborough, A. Wu, J. Tellez, C. Ekstrand, E. Lorentz, T. Ellchuk, P. Babyn  
430 & R. Borowsky. 2016. Presurgical language mapping in epilepsy: Using fMRI of reading  
431 to identify functional reorganization in a patient with long-standing temporal lobe  
432 epilepsy. *Epilepsy Behav Case Rep* 5.6-10.
- 433 Grouiller, F., J. Jorge, F. Pittau, W. van der Zwaag, G. R. Iannotti, C. M. Michel, S. Vulliemoz,  
434 M. I. Vargas & F. Lazeyras. 2016. Presurgical brain mapping in epilepsy using  
435 simultaneous EEG and functional MRI at ultra-high field: feasibility and first results.  
436 *MAGMA* 29.605-16.

- 437 Hakyemez, B, C Erdogan, N Yildirim, I Bora, A Bekar & M Parlak. 2006. Functional MRI in  
438 patients with intracranial lesions near language areas. *The neuroradiology journal* 19.306-  
439 12.
- 440 Harrington, Greg S, MH Buonocore & S Tomaszewski Farias. 2006. Intrasubject reproducibility  
441 of functional MR imaging activation in language tasks. *American Journal of*  
442 *Neuroradiology* 27.938-44.
- 443 Jansen, A., R. Menke, J. Sommer, A. F. Forster, S. Bruchmann, J. Hempleman, B. Weber & S.  
444 Knecht. 2006. The assessment of hemispheric lateralization in functional MRI--  
445 robustness and reproducibility. *NeuroImage* 33.204-17.
- 446 Khateb, Asaid, Marie-Dominique Martory, Jean-Marie Annoni, François Lazeyras, Nicolas de  
447 Tribolet, Alan J Pegna, Eugène Mayer, Christoph M Michel & Mohamed L Seghier.  
448 2004. Transient crossed aphasia evidenced by functional brain imagery. *Neuroreport*  
449 15.785-90.
- 450 Knecht, S, A Jansen, A Frank, J Van Randenborgh, J Sommer, M Kanowski & HJ Heinze. 2003.  
451 How atypical is atypical language dominance? *NeuroImage* 18.917-27.
- 452 Kononen, M., N. Tamsi, L. Saisanen, S. Kemppainen, S. Maatta, P. Julkunen, L. Jutila, M. Aikia,  
453 R. Kalviainen, E. Niskanen, R. Vanninen, P. Karjalainen & E. Mervaala. 2015. Non-  
454 invasive mapping of bilateral motor speech areas using navigated transcranial magnetic  
455 stimulation and functional magnetic resonance imaging. *J Neurosci Methods* 248.32-40.
- 456 Kośła, Katarzyna, Bartosz Bryszewski, Dariusz Jaskólski, Nina Błasiak-Kołacińska, Ludomir  
457 Stefańczyk & Agata Majos. 2015. Reorganization of language areas in patient with a  
458 frontal lobe low grade glioma–fMRI case study. *Polish journal of radiology* 80.290.
- 459 Kragel, P. A., X. Han, T. E. Krainak, P. J. Gianaros & T. D. Wager. 2021. Functional MRI Can  
460 Be Highly Reliable, but It Depends on What You Measure: A Commentary on Elliott et  
461 al. (2020). *Psychol Sci.*956797621989730.
- 462 Kriegeskorte, Nikolaus, Rainer Goebel & Peter Bandettini. 2006. Information-based functional  
463 brain mapping. *Proceedings of the National Academy of Science* 103.3863-68.
- 464 Laumann, T. O., E. M. Gordon, B. Adeyemo, A. Z. Snyder, S. J. Joo, M. Y. Chen, A. W.  
465 Gilmore, K. B. McDermott, S. M. Nelson, N. U. Dosenbach, B. L. Schlaggar, J. A.  
466 Mumford, R. A. Poldrack & S. E. Petersen. 2015. Functional System and Areal  
467 Organization of a Highly Sampled Individual Human Brain. *Neuron* 87.657-70.
- 468 Lee, D., S. J. Swanson, D. S. Sabsevitz, T. A. Hammeke, F. Scott Winstanley, E. T. Possing & J.  
469 R. Binder. 2008. Functional MRI and Wada studies in patients with interhemispheric  
470 dissociation of language functions. *Epilepsy Behav* 13.350-6.
- 471 Liberati, Alessandro, Douglas G Altman, Jennifer Tetzlaff, Cynthia Mulrow, Peter C Gøtzsche,  
472 John PA Ioannidis, Mike Clarke, Philip J Devereaux, Jos Kleijnen & David Moher. 2009.  
473 The PRISMA statement for reporting systematic reviews and meta-analyses of studies  
474 that evaluate health care interventions: explanation and elaboration. *PLoS medicine*  
475 6.e1000100.
- 476 Mahowald, K. & E. Fedorenko. 2016. Reliable individual-level neural markers of high-level  
477 language processing: A necessary precursor for relating neural variability to behavioral  
478 and genetic variability. *NeuroImage* 139.74-93.
- 479 Maldjian, Joseph A, Paul J Laurienti, Lance Driskill & Jonathan H Burdette. 2002. Multiple  
480 reproducibility indices for evaluation of cognitive functional MR imaging paradigms.  
481 *American Journal of Neuroradiology* 23.1030-37.

- 482 McGonigle, David J. 2012. Test–retest reliability in fMRI: or how I learned to stop worrying and  
483 love the variability. *NeuroImage* 62.1116-20.
- 484 McGonigle, David J, Alistair M Howseman, Balwinder S Athwal, Karl J Friston, RSJ  
485 Frackowiak & Andrew P Holmes. 2000. Variability in fMRI: an examination of  
486 intersession differences. *NeuroImage* 11.708-34.
- 487 Nettekoven, C., N. Reck, R. Goldbrunner, C. Grefkes & C. Weiss Lucas. 2018. Short- and long-  
488 term reliability of language fMRI. *Neuroimage* 176.215-25.
- 489 Noble, Stephanie, Dustin Scheinost & R Todd Constable. 2019. A decade of test-retest reliability  
490 of functional connectivity: A systematic review and meta-analysis. *NeuroImage*  
491 203.116157.
- 492 Nosek, Brian A. & Daniël Lakens. 2014. Registered reports: A method to increase the credibility  
493 of published results. *Social Psychology* 45.137-41.
- 494 Otzenberger, H., D. Gounot, C. Marrer, I. J. Namer & M. N. Metz-Lutz. 2005. Reliability of  
495 individual functional MRI brain mapping of language. *Neuropsychology* 19.484-93.
- 496 Paek, E. J. Murray L. L. Newman S. D. Kim D. J. 2019. Test-retest reliability in an fMRI study  
497 of naming in dementia. *Brain & Language* 191.31-45.
- 498 Partovi, S, B Jacobi, N Rapps, L Zipp, S Karimi, F Rengier, JK Lyo & C Stippich. 2012. Clinical  
499 standardized fMRI reveals altered language lateralization in patients with brain tumor.  
500 *American Journal of Neuroradiology* 33.2151-57.
- 501 Peck, Kyung K, Michelle Bradbury, Nicole Petrovich, Bob L Hou, Nicole Ishill, Cameron  
502 Brennan, Viviane Tabar & Andrei I Holodny. 2009. Presurgical evaluation of language  
503 using functional magnetic resonance imaging in brain tumor patients with previous  
504 surgery. *Neurosurgery* 64.644-53.
- 505 Peelle, Jonathan E. 2012. The hemispheric lateralization of speech processing depends on what  
506 “speech” is: a hierarchical perspective. *Frontiers in Human Neuroscience* 6.
- 507 Polczynska, M, C. F. Benjamin, K. Japardi, A. Frew & S. Y. Bookheimer. 2016. Language  
508 system organization in a quadrilingual with a brain tumor: Implications for understanding  
509 of the language network. *Neuropsychologia* 86.167-75.
- 510 Polczynska, M, K. Japardi & S. Y. Bookheimer. 2017. Lateralizing language function with pre-  
511 operative functional magnetic resonance imaging in early proficient bilingual patients.  
512 *Brain Lang* 170.1-11.
- 513 Poldrack, R. A., C. I. Baker, J. Durnez, K. J. Gorgolewski, P. M. Matthews, M. R. Munafo, T. E.  
514 Nichols, J. B. Poline, E. Vul & T. Yarkoni. 2017. Scanning the horizon: towards  
515 transparent and reproducible neuroimaging research. *Nature Reviews Neuroscience*  
516 18.115-26.
- 517 Poldrack, Russell A. & Krzysztof J. Gorgolewski. 2014. Making big data open: data sharing in  
518 neuroimaging. *Nature Neuroscience* 17.1510-17.
- 519 Poldrack, Russell A., Timothy O. Laumann, Oluwasanmi Koyejo, Brenda Gregory, Ashleigh  
520 Hover, Mei-Yen Chen, Krzysztof J. Gorgolewski, Jeffrey Luci, Sung Jun Joo, Ryan L.  
521 Boyd, Scott Hunicke-Smith, Zack Booth Simpson, Thomas Caven, Vanessa Sochat,  
522 James M. Shine, Evan Gordon, Abraham Z. Snyder, Babatunde Adeyemo, Steven E.  
523 Petersen, David C. Glahn, D. Reese McKay, Joanne E. Curran, Harald H. H. Göring,  
524 Melanie A. Carless, John Blangero, Robert Dougherty, Alexander Leemans, Daniel A.  
525 Handwerker, Laurie Frick, Edward M. Marcotte & Jeanette A. Mumford. 2015. Long-  
526 term neural and physiological phenotyping of a single human. *Nature Communications*  
527 6.8885.

- 528 Poldrack, Russell, Deanna M Barch, Jason Mitchell, Tor Wager, Anthony D Wagner, Joseph T  
529 Devlin, Chad Cumba, Oluwasanmi Koyejo & Michael Milham. 2013. Toward open  
530 sharing of task-based fMRI data: the OpenfMRI project. *Frontiers in Neuroinformatics*  
531 7.12.
- 532 Poldrack, Russell, Aniket Kittur, Donald Kalar, Eric Miller, Christian Seppa, Yolanda Gil, D.  
533 Parker, Fred Sabb & Robert Bilder. 2011. The Cognitive Atlas: Toward a Knowledge  
534 Foundation for Cognitive Neuroscience. *Frontiers in Neuroinformatics* 5.
- 535 Pouratian, Nader, Susan Y Bookheimer, David E Rex, Neil A Martin & Arthur W Toga. 2002.  
536 Utility of preoperative functional magnetic resonance imaging for identifying language  
537 cortices in patients with vascular malformations. *Journal of Neurosurgery* 97.21-32.
- 538 Rau, S. Fesl G. Bruhns P. Havel P. Braun B. Tonn J. C. Ilmberger J. 2007. Reproducibility of  
539 activations in broca area with two language tasks: A functional MR imaging study.  
540 *American Journal of Neuroradiology* 28.1346-53.
- 541 Röder, Brigitte, Oliver Stock, Siegfried Bien, Helen Neville & Frank Rösler. 2002. Speech  
542 processing activates visual cortex in congenitally blind humans. *European Journal of*  
543 *Neuroscience* 16.930-36.
- 544 Roland, J. L., C. D. Hacker, J. D. Breshears, C. M. Gaona, R. E. Hogan, H. Burton, M. Corbetta  
545 & E. C. Leuthardt. 2013. Brain mapping in a patient with congenital blindness - a case for  
546 multimodal approaches. *Front Hum Neurosci* 7.431.
- 547 Ruff, I. M., N. M. Petrovich Brennan, K. K. Peck, B. L. Hou, V. Tabar, C. W. Brennan & A. I.  
548 Holodny. 2008. Assessment of the language laterality index in patients with brain tumor  
549 using functional MR imaging: effects of thresholding, task selection, and prior surgery.  
550 *AJNR Am J Neuroradiol* 29.528-35.
- 551 Rutten, G. J., N. F. Ramsey, P. C. van Rijen, H. J. Noordmans & C. W. van Veelen. 2002.  
552 Development of a functional magnetic resonance imaging protocol for intraoperative  
553 localization of critical temporoparietal language areas. *Ann Neurol* 51.350-60.
- 554 Seghier, M. L., F. Lazeyras, A. J. Pegna, J. M. Annoni, I. Zimine, E. Mayer, C. M. Michel & A.  
555 Khateb. 2004. Variability of fMRI activation during a phonological and semantic  
556 language task in healthy subjects. *Hum Brain Mapp* 23.140-55.
- 557 Seghier, Mohamed L. 2008. Laterality index in functional MRI: methodological issues. *Magnetic*  
558 *resonance imaging* 26.594-601.
- 559 Sierpowska, J., A. Gabarros, P. Ripolles, M. Juncadella, S. Castaner, A. Camins, G. Plans & A.  
560 Rodriguez-Fornells. 2013. Intraoperative electrical stimulation of language switching in  
561 two bilingual patients. *Neuropsychologia* 51.2882-92.
- 562 Simmons, Joseph P., Leif D. Nelson & Uri Simonsohn. 2011. False-positive psychology:  
563 Undisclosed flexibility in data collection and analysis allows presenting anything as  
564 significant. *Psychological Science* 22.1359-66.
- 565 Smith, Stephen M, Christian F Beckmann, Narender Ramnani, Mark W Woolrich, Peter R  
566 Bannister, Mark Jenkinson, Paul M Matthews & David J McGonigle. 2005. Variability in  
567 fMRI: a re-examination of inter-session differences. *Human Brain Mapping* 24.248-57.
- 568 Tailby, Chris, David F Abbott & Graeme D Jackson. 2017. The diminishing dominance of the  
569 dominant hemisphere: language fMRI in focal epilepsy. *NeuroImage: Clinical* 14.141-50.
- 570 Team, R Core. 2013. R: A language and environment for statistical computing: Vienna, Austria.
- 571 Tie, Y., R. O. Suarez, S. Whalen, A. Radmanesh, I. H. Norton & A. J. Golby. 2009. Comparison  
572 of blocked and event-related fMRI designs for pre-surgical language mapping.  
573 *Neuroimage* 47 Suppl 2.T107-15.

## Single-subject neuroimaging studies of language processing

- 574 Tomasino, B., D. Marin, C. Canderan, M. Maieron, R. Budai, F. Fabbro & M. Skrap. 2014.  
575 Involuntary switching into the native language induced by electrocortical stimulation of  
576 the superior temporal gyrus: a multimodal mapping study. *Neuropsychologia* 62.87-100.
- 577 Van Horn, John Darrell, Scott T Grafton & Michael B Miller. 2008. Individual variability in  
578 brain activity: a nuisance or an opportunity? *Brain imaging and behavior* 2.327.
- 579 Voyvodic, James T. 2012. Reproducibility of single-subject fMRI language mapping with  
580 AMPLE normalization. *Journal of magnetic resonance imaging* 36.569-80.
- 581 Wilson, S. M., A. Bautista, M. Yen, S. Lauderdale & D. K. Eriksson. 2017. Validity and  
582 reliability of four language mapping paradigms. *Neuroimage Clin* 16.399-408.
- 583 Wilson, S. M., M. Yen & D. K. Eriksson. 2018. An adaptive semantic matching paradigm for  
584 reliable and valid language mapping in individuals with aphasia. *Hum Brain Mapp*  
585 39.3285-307.
- 586 Woo, C. W., L. J. Chang, M. A. Lindquist & T. D. Wager. 2017. Building better biomarkers:  
587 brain models in translational neuroimaging. *Nat Neurosci* 20.365-77.
- 588 Yen, M. DeMarco A. T. Wilson S. M. 2019. Adaptive paradigms for mapping phonological  
589 regions in individual participants. *NeuroImage* 189.368-79.
- 590