Investigating Attentional Allocation With Eye Tracking During Category Learning in People With Aphasia

Sofia Vallila-Robter and Brendan Czupryna

Studies have identified deficits in attention in individuals with aphasia in language and nonlanguage tasks. Attention may play a role in the construction and use of language, as well as in learning and the process of rehabilitation, yet the role of attention on rehabilitation is not fully understood. To improve the understanding of attention and learning in aphasia, this study replicated an experiment that utilized category learning to examine attentional allocation. Ten individuals with aphasia subsequent to left hemisphere stroke and 20 age-matched controls completed a computer-based category learning task while eye gaze data were collected using an eye tracker. Stimulus items comprised 4 features that differed in the reliability with which they predicted category membership (referred to as their diagnosticity). In this study, no differences were observed between individuals with aphasia and control participants on behavioral measures of accuracy and response time, though accuracies overall were lower than those of prior studies examining this task in young adults. Eye gaze data demonstrated that over the course of training, controls and individuals with aphasia learned to reduce the number of looks to the feature of lowest diagnosticity, suggestive of optimized attentional allocation. Eye gaze patterns, however, did not show increased looking or look times to all features of highest diagnosticity, which has been seen in young adults. Older adults and individuals with aphasia may benefit from additional processing time or additional trials during category learning to optimize attention and behavioral accuracy. Findings are relevant to consider in clinical settings where visual stimuli are presented as instructional, supporting, and/or compensatory tools. Key words: aphasia, attentional allocation, category learning, eye tracking

A PHASIA has been historically described as a deficit of language processing and expression. Although linguistic inefficiencies remain the primary area of weakness,

The authors thank the Christopher Norman Fund for contributing compensation to research participants. They also thank Emmaleigh Loyer for ber contributions to data analysis.

The authors have indicated that they have no financial and no nonfinancial relationships to disclose.

Corresponding Author: Sofia Vallila-Robter, PbD, Department of Communication Sciences and Disorders, MGH Institute of Health Professions, 36 First Ave, Boston, MA 02129 (svallilarobter@mgbibp.edu).

DOI: 10.1097/TLD.000000000000206

researchers have increasingly recognized the contributions of other nonlinguistic cognitive processes in the profile of a person with aphasia, including deficits in attention (Marshall, Basilakos, & Love-Myers, 2013; Murray, 1999, 2012), executive functioning (Fucetola, Connor, Strube, & Corbetta, 2009; Glosser & Goodglass, 1990; Lesniak, Bak, Czepiel, Seniow, & Czlonkowska, 2008; Purdy, 2002), memory (Burgio & Basso, 1997; Potagas, Kasselimis, & Evdokimidis, 2011), and visuospatial skills (Helm-Estabrooks, 2002). Language therapy utilizes principles of linguistic organization and retrieval to restore and rebuild language access. Nonlinguistic cognitive processes not only shape clients' engagement with a task but also impact the linguistic processing itself.

Author Affiliation: Department of Communication Sciences and Disorders, MGH Institute of Health Professions, Boston, Massachusetts.

Of particular relevance to this study is attention. Studies have demonstrated that individuals with aphasia often present with attention deficits relative to controls (Erickson, Goldinger, & LaPointe, 1996; Heuer & Hallowell, 2015; Hunting-Pompon, Kendall, & Moore, 2011; Laures, 2005; Laures, Odell, & Coe, 2003; Murray, 2000, 2012; Murray, Holland, & Beeson, 1997; Peach, Newhoff, & Rubin, 1993; Petry, Crosson, Rothi, Bauer, & Schauer, 1994; Robin & Rizzo, 1989; Tseng, McNeil, & Milenkovic, 1993; Villard & Kiran, 2015; 2017) and, furthermore, that increased individual variability in attention abilities may exist relative to controls (Villard & Kiran, 2015, 2018). Reduced attention has been measured in contexts that require language (Murray, 2000, 2012; Murray et al., 1997; Petry et al., 1994; Tseng et al., 1993) and also in nonlanguage tasks (Erickson et al., 1996; Hunting-Pompon et al., 2011; Laures, 2005; Laures et al., 2003; Peach et al., 1993; Robin & Rizzo, 1989; Villard & Kiran, 2015, 2017, 2018). Deficits in attention have been hypothesized by some to be at the root of lexical retrieval deficits (Hula & McNeil, 2008; Hula, McNeil, & Sung, 2007; McNeil, Odell, & Tseng, 1991; Silkes, McNeil, & Drton, 2004).

In addition to impacting language function and use, attention plays a critical role in focusing on tasks in rehabilitation to support the process of recovery (Villard, 2017; Villard & Kiran, 2015, 2018). Some study results demonstrate that individuals with greater attention abilities produce greater outcomes (Lambon Ralph, Snell, Fillingham, Conroy, & Sage, 2010; Mysiw, Beegan, & Gatens, 1989; Robertson, Ridgeway, Greenfield, & Parr, 1997). This may be in part due to the fact that aphasia intervention requires auditory and visual attention as individuals with aphasia listen to auditory stimuli and/or clinician feedback and instruction while looking at pictures or written prompts (Villard, 2017). Attending to and integrating these domains are critical for rehabilitation to progress.

In addition to deficits in attention surfacing in individuals with aphasia, impairments in learning have been observed in domains such as category learning (Vallila-Rohter & Kiran, 2013, 2015), explicit sequence learning (Schuchard & Thompson, 2014), auditory (Goschke, Friederici, Kotz, & Kampen, 2001) and visual artificial grammar learning (Christiansen, Kelly, Shillcock, & Greenfield, 2010), and word and sentence learning (Ettlinger & Moffett, 1970; Grossman & Carey, 1987; Gupta, Martin, Abbs, Schwartz, & Lipinksi, 2006; Martin & Saffran, 1999). Other study findings have demonstrated intact learning in individuals with aphasia, with people with aphasia showing the ability to learn novel words (Kelly & Armstrong, 2009; Tuomiranta et al., 2011; Tuomiranta, Grönroos, Martin, & Laine, 2014; Tuomiranta, Rautakoski, Rinne, Martin, & Laine, 2012), artificial grammars (Glass, Gazzaniga, Premack, 1973; Schuchard & Thompson, 2017), or sequences implicitly (Schuchard & Thompson, 2014). Contradicting findings across studies that examine learning suggests that variability arises in aphasia and that learning within this population is sensitive to stimulus manipulations, meriting further study. Sensitivity to stimulus manipulations and learning conditions is not unique to aphasia and has been observed in other clinical populations such as Parkinson's disease, Huntington's disease, and amnesia (Ashby, Noble, Filoteo, Waldron, & Ell, 2003; Koenig, Smith, Moore, Glosser, & Grossman, 2007; Knowlton, Ramus, & Squire, 1992; Shohamy, Myers, Onlaor, & Gluck, 2004).

In a study by Vallila-Rohter and Kiran (2013), individuals with aphasia were asked to learn to categorize animal exemplars into two categories based on the probabilistic overlap of visual features with two prototypes. Results indicated that although some people with aphasia showed evidence of learning, many did not learn and, as a group, they performed less accurately than an age-matched control group. Accuracy of categorization did not correlate with language and/or cognitive ability. A follow-up study (Vallila-Rohter & Kiran, 2015) examined the strategies utilized during learning. Strategy analyses revealed that

many people with aphasia used suboptimal strategies during learning, attending to a single stimulus dimension rather than basing responses on multiple stimulus dimensions. Poor performance may have resulted from impaired learning mechanisms but may also have been related to a reduced ability to examine and process multidimensional visual stimuli.

Thus, building upon previously documented deficits in attention and learning in aphasia, the current study replicated an experiment first conducted by Rehder and Hoffman (2005) to examine visual attention in individuals with aphasia and age-matched controls through the course of a category-learning task. In the study by Rehder and Hoffman (2005), young adults sorted stimuli into two categories while eye-tracking data were collected. Category membership was based on the combination of four features-head, wing, tail, and foot-arranged around a central rectangular body. Importantly, two of the features were highly diagnostic, meaning that they were highly predictive of category membership. In contrast, the remaining features were classified as being of medium and low diagnosticity, indicating that they less consistently predicted category membership. Participants showed a pattern of optimized attentional allocation as learning progressed. Although all features were fixated at the onset of learning, over time, fixations increased to features of high diagnosticity and decreased to the feature of lowest diagnosticity reflecting changes in attentional allocation through the course of learning.

In this study, which replicated Rehder and Hoffman's (2005) methods, we hypothesized that older adults would show the patterns previously observed, corresponding to an increase in frequency and duration of eye gazes to features of high predictive value alongside decreased eye gazes to features of low predictive value. In contrast, we hypothesized that people with aphasia would not show optimized attentional allocation and would rather either fixate on all features equally throughout learning or exhibit a tendency to fixate on one or two dimensions only.

METHODS

Participants

Thirty participants completed the study. Ten were right-handed individuals with aphasia (seven men, three women) in the chronic stage of recovery from a single left hemisphere stroke. Participants with aphasia ranged in age from 36 to 63 years (M = 54.5, SD = 8.5) and averaged 72 months poststroke (range: 6-97 months). Individuals with aphasia completed the Western Aphasia Battery-Revised (WAB-R; Kertesz, 2006) to classify aphasia and completed the Cognitive Linguistic Quick Test (CLQT; Helm-Estabrooks, 2001) to characterize nonverbal cognitive processing and visual perceptive skills (see Table 1). The average WAB Aphasia Quotient of enrolled participants was 86.79 (SD = 8.6), with nine subjects classified as anomic and one subject classified as having conduction aphasia. The WAB scores range from 0 to 100, with 0 indicating the most severe language impairments and scores 93.8 and above corresponding to language within normal limits.

Twenty participants were right-handed controls (10 men, 10 women) ranging from 51 to 76 years of age (M = 63.1, SD = 7.0) with no history of neurological disease, psychological disorder, or developmental speech, language, or learning disabilities. We recruited more controls than individuals with aphasia, as in addition to comparing performance between individuals with aphasia and controls, we aimed to compare results with those from the original study by Rehder and Hoffman (2005). Rehder and Hoffman (2005) studied attentional allocation in undergraduate students. We were interested in comparing results from older adults in the current sample with previously published results collected from undergraduate participants. Control participants completed the Mini-Mental State Examination to screen for cognitive decline and had to be within the normal range of cognition (24 and above) to be eligible. Cognitive Linguistic Quick Test measures of cognitive ability were also obtained.

| | WAB AQ | Age | Attn | Mem | EF | vs |
|------------------------|--------|-----|------|-----|----|-----|
| Control 1 | | 64 | 200 | 185 | 33 | 97 |
| Control 2 | | 67 | 205 | 171 | 31 | 99 |
| Control 3 | | 65 | 187 | 164 | 26 | 87 |
| Control 4 | | 65 | 186 | 163 | 24 | 87 |
| Control 5 | | 64 | 205 | 167 | 33 | 101 |
| Control 6 | | 57 | 202 | 166 | 32 | 98 |
| Control 7 | | 58 | 205 | 170 | 30 | 99 |
| Control 8 | | 62 | 195 | 172 | 30 | 91 |
| Control 9 | | 52 | 201 | 153 | 31 | 91 |
| Control 10 | | 71 | 194 | 170 | 27 | 91 |
| Control 11 | | 64 | 204 | 165 | 31 | 100 |
| Control 12 | | 72 | 186 | 171 | 31 | 94 |
| Control 13 | | 75 | 198 | 169 | 29 | 91 |
| Control 14 | | 75 | 209 | 178 | 32 | 101 |
| Control 15 | | 50 | 208 | 171 | 31 | 102 |
| Control 16 | | 54 | 194 | 163 | 32 | 93 |
| Control 17 | | 61 | 208 | 172 | 32 | 102 |
| Control 18 | | 58 | 197 | 156 | 29 | 93 |
| Control 19 | | 59 | 210 | 185 | 33 | 100 |
| Control 20 | | 59 | 200 | 166 | 31 | 97 |
| Person with Aphasia 1 | 95.6 | 61 | 207 | 170 | 30 | 101 |
| Person with Aphasia 2 | 93.4 | 58 | 201 | 152 | 27 | 95 |
| Person with Aphasia 3 | 85.8 | 53 | 200 | 143 | 27 | 102 |
| Person with Aphasia 4 | 89.5 | 60 | 187 | 125 | 28 | 92 |
| Person with Aphasia 5 | 92.4 | 36 | 198 | 158 | 30 | 96 |
| Person with Aphasia 6 | 85.9 | 63 | 191 | 117 | 27 | 96 |
| Person with Aphasia 7 | 80.9 | 53 | 194 | 138 | 28 | 98 |
| Person with Aphasia 8 | 94.3 | 63 | 206 | 158 | 30 | 104 |
| Person with Aphasia 9 | 73.2 | 47 | 194 | 144 | 27 | 97 |
| Person with Aphasia 10 | 74.9 | 50 | 198 | 147 | 25 | 98 |

Table 1. Participant characteristics

Note. WAB AQ stands for the Western Aphasia Battery Aphasia Quotient, a measure of severity of aphasia. Attn, Mem, EF, and VS stand for attention, memory, executive functions, and visuospatial skills, respectively, as measured on the Cognitive Linguistic Quick Test (CLQT).

Because of the visual nature of the task and the eye-tracking paradigm, participants could not miss more than two symbols per quadrant on the CLQT Symbol Cancellation Task to be eligible. Participants who wore bifocals or hard lenses incompatible with the eyetracking system were ineligible to participate.

Experimental setup

Participants were seated approximately 50 cm from a 53- \times 30-cm computer screen with 1,920 \times 1,080 resolution and 120 Hz

refresh rate. Eye fixations were captured using an SR Eyelink 1000 Plus Eye Tracker (SR Research, Canada) with a sampling rate of 1,000 Hz and a noise-limited spatial resolution better than 0.01° . A temporal threshold of 100 ms was used to determine a fixation. Monocular data were gathered with head position stabilized via chin rest. Images presented on the computer screen occupied a 12° of visual angle. At the start of testing, eye gaze was calibrated using nine-point calibration and validation. Drift corrects offering the opportunity for recalibration were programmed throughout the experiment, specifically at the beginning of the training phase, between Blocks 10 and 11 of training, and prior to the testing block.

Stimuli

As noted previously, the experimental task was a replication of Rehder and Hoffman's (2005) study. Experimental stimuli were provided by Dr. Rehder and consisted of 16 color line drawings of "insects," each with four features (head, tail, foot, and wing) arranged around a central rectangular body. Each feature was approximately 4° in width and height and had two possible variations, coded as 1 or 0. For example, the head could be circular or an elongated pentagon, the wing could be a triangle or crescent (see Figure 1). Each animal was labeled by its four binary dimensions (e.g., 0100, 0110, 1110).

Training stimuli consisted of nine exemplars drawn from the 16 total possible stimuli/ feature combinations (five Category A trained items: 1110, 1010, 1011, 1101, 0111; four Category B trained items: 1100, 0110, 0001, 0000). The selection of nine items for training is rooted in prior category-learning work designed to distinguish between prototype and exemplar theories of learning (see Nosofsky & Zaki, 2002, for more details). Prototype theories suggest that category decisions are based on the similarities between a stimulus item and the most central or typical item of a category (the category prototype). Even if not presented in training, a category prototype is thought to be acquired through gradual exposure to category regularities. In contrast, exemplar theories suggest that category decisions are driven by the similarity of one stimulus item seen in training to others. The 5-4 category structure produces within-category items that differ in their expected categorization accuracies based on prototype versus exemplar theories. These predictions are best elucidated by focusing on categorization predictions from Category A stimulus items: 1110 and 1010. From the perspective of prototype theory, stimulus 1110 (of Category A) is very similar to prototype A (1111), differing on only one feature dimension. In contrast, stimulus 1010 differs from the prototype by two feature dimensions. Prototype theories predict a higher A categorization of item 1110 than of item 1010. In contrast, exemplar theories suggest that categorization is based on the similarity of training items with other within-category trained items, rather than with an imagined prototype. From the perspective of exemplar theory, therefore, stimulus 1010 shares three features with two Category A stimulus items (1110 and 1011) and a maximum of two features with Category B items. In contrast, stimulus 1110 shares three features with one Category A stimulus (1010) and with two Category B stimuli (1100 and 0110). Thus, exemplar theorists predict higher A categorization of item 1010 than item 1110. In this manner, the 5-4 category structure allows for comparisons of



Figure 1. Sample training trial showing a stimulus animal with four features: head (medium diagnosticity), tail (low diagnosticity), wing (high diagnosticity), and foot (high diagnosticity). Boxes identify areas of interest and indicate that eye gaze data from this interest period were analyzed.

prototype versus exemplar approaches to learning. In this study, we replicated this design to remain consistent with Rehder's original study, but analyses of learning strategy are beyond the scope of the this manuscript and underpowered.

As in the study by Rehder and Hoffman (2005), the features wing and foot were highly diagnostic, meaning that the feature dimension independently predicted category membership on 77% of trials. The head feature predicted category assignment on 66% of trials and was considered of medium diagnosticity. The tail had low diagnosticity and predicted category membership independently on 55% of trials. Two task versions were created to account for the fact that five animals were trained in one category and four in the other. In one task version, the triangular wing was coded to be most frequently associated with a correct response of A. In the alternate task version, the crescent wing was most frequently associated with a correct response of A. Task versions were counterbalanced across participants.

Procedures

The experimental task was first explained verbally to participants using a script and accompanying printed handouts with pictures. Participants were given the opportunity to ask questions and clarification. Participants then received computer-based instructions that allowed them to practice indicating responses via button press and familiarized them with the timing of the task and feedback presentation. Participants were told that they must classify animals as belonging to one of two categories by pressing the 1 or 2 key, respectively. Because many people with aphasia who experience a left hemisphere stroke have right-sided weakness, all participants were instructed to make responses with their left hand. Auditory feedback was provided after each trial in the learning phase in the form an ascending two-tone chime (correct), or descending two-tone chime (incorrect). Participants heard these sounds several times before the task to set volume levels and ensure recognition and comprehension of feedback signals.

The task was composed of two phases: training and testing. In training, each trial started with a 1,000-ms fixation cross in the center of the screen. A stimulus item then appeared on screen and participants had up to 3,000 ms to make a button press to indicate the item's category membership. A time limit was imposed to introduce relative consistency to the amount of time spent looking at stimuli. Without response limitations, some participants might have studied stimuli for much longer than others. Auditory feedback was provided immediately via headphones and the stimulus item remained on screen for an additional 4,000 ms after feedback (see Figure 1). All nine training stimuli were presented in random order within each block. Training concluded when participants either completed 21 blocks (189 trials) or reached criterion. Criterion was defined as completing two consecutive blocks without error.

In the testing phase that immediately followed training, participants categorized trained items and the seven items not previously seen in training (1001, 1000, 1111, 0010, 0101, 0011, 0100) to probe generalization of learning to novel items. The 16-item test set was presented twice in randomized order. Once again, a fixation point appeared in the middle of the screen for 1,000 ms, after which a stimulus appeared. Participants had unlimited time to respond and the subsequent trial was triggered after a button press response was made. No feedback was provided in testing.

Dependent measures

Behavioral data were collected on accuracy, reaction time, and number of blocks to reach criterion (two consecutive blocks without error). Two eye-tracking measures were derived on the basis of fixation data within areas of interest (AOIs). Areas of interest were drawn around the four features of the stimuli. These AOIs were polygons encompassing the total area of the feature on the screen. Shape and size of each AOI varied to capture the associated feature but were controlled for distance from the central body. Eye-tracking analyses were conducted only over the interest period from onset of stimulus to response in order to better understand what features participants inspected to make responses. Two dependent measures were derived and averaged across trials in a block: number of observations and proportion log fixation time. Number of observations indicated the number of times a dimension was fixated during a trial. For this measure, two sequential fixations to a single dimension were aggregated into a single "observation." To compute proportion log fixation time, log fixation time was first measured as the total number of milliseconds that a dimension was fixated by trial and averaged over block. Because of the nonnormal nature of fixation time data, fixation times were log-transformed. Then, eight average log fixation times were calculated for each participant to compare fixations to features of varying diagnosticity and to make comparisons early and late in training: Head early log fixation, Tail early log fixation, Wing early log fixation and Foot early log fixation time, Head late log fixation, Tail late log fixation, Wing late log fixation, and Foot late log fixation time. Averages for early training were computed by averaging data from training Blocks 1, 2, 3, and 4 (early). Data from training Blocks 18, 19, 20, and 21 contributed to averages for late phases of training. Data for each participant were converted to a proportion log fixation based on the average log fixation time to a specific feature divided by the total log fixation time to all features (Figure 2).

Examination of individual participant data revealed that many participants focused on one feature of high diagnosticity and not the



Figure 2. Number of observations and proportion log fixation time results for control participants (left) and individuals with aphasia across 21 blocks of learning. PWA = participants with aphasia.

other. Although the two features of high diagnosticity were equally predictive of category membership, participants exhibited a tendency to fixate on one of the two, and the feature of focus varied across participants. To account for this discrepancy, in a manner similar to that implemented by Arbel, Feeley, and Xinyi (2019), for each participant, the feature with the higher fixation probability and log fixation time was coded as HighHigh and the other as HighLow. For each participant, one feature was coded as HighHigh on the basis of performance across all blocks. The feature wing was coded as HighHigh for seven controls and two individuals with aphasia.

Analyses

Statistical analyses were conducted in R (R Core Team, 2014). For each participant, three average accuracies were calculated: early training (average accuracy over Blocks 1-4), late training (average accuracy over Blocks 18-21), and test accuracy. Average early training response times (Blocks 1-4) and late training response times (Blocks 18-21) were also determined. Accuracy and response time were evaluated during early and late training phases as eye gaze data focus on these blocks. A series of 2 (Phase: early, late) \times 2 (Group: controls, individuals with aphasia) analyses of variance (ANOVAs) were conducted to evaluate for differences in accuracy and response time.

To evaluate eye gaze behaviors, linear mixed-effects models were utilized using R package lme4 (Bates, Maechler, Bolker, & Walker, 2015) to examine dependent variables with Group (people with aphasia or controls), Diagnosticity (HighHigh, HighLow, Medium, Low), and Phase (early, late) as fixed effects, and Participant as a random effect. As Group and Diagnosticity data were categorical, it was not appropriate to model random slopes. We evaluated whether there were any significant interactions of diagnosticity and phase to gain insights into changing attentional allocation over the course of learning. Our model for the dependent variable number of observations was as follows: model = lmer (NumObservations ~ Group + DiagClass + Phase + Group*DiagClass*Phase + (1|Participant). R package lmerTest (Kuznetsova, Brockoff, & Christensen, 2017) was used to conduct ANOVAs evaluating models. For proportion log fixation time, package glmmadmb (Skaug, Fournier, Nielsen, Magnusson, & Bolker, 2011) was used to account for the use of continuous proportions as the dependent variable.

RESULTS

Behavioral results

No participants reached criterion of two consecutive blocks without error and therefore all participants completed all training trials prior to testing. Average accuracy early in training, late in training, and in testing is reported in Table 2. The 2 (Phase: early, late) \times 2 (Group: controls, individuals with aphasia) ANOVA evaluating training accuracy produced a significant effect of phase *F*(1, 57) = 11.32, *p* = .001, with scores increasing from

Table 2. Average accuracies and responsetimes across groups and training phases

| | Mean | SD | |
|----------|---------|--------|--|
| Accuracy | | | |
| CN | | | |
| Early | 51.81% | 9.68% | |
| Late | 63.61% | 12.29% | |
| Test | 61.42% | 11.18% | |
| PWA | | | |
| Early | 57.50% | 9.89% | |
| Late | 62.50% | 10.90% | |
| Test | 62.60% | 12.94% | |
| RT (ms) | | | |
| CN | | | |
| Early | 1491.59 | 187.51 | |
| Late | 1496.30 | 346.92 | |
| PWA | | | |
| Early | 1433.68 | 299.80 | |
| Late | 1563.96 | 387.14 | |

Note. CN = controls; PWA = participants with aphasia; RT = response time.

early to late training. There was no significant effect of group F(1, 57) = 0.58, p =.45, demonstrating that overall accuracy was similar for controls and individuals with aphasia. The interaction of group and phase was nonsignificant, F(1, 57) = 1.37, p = .24. A 2 (Phase) \times 2 (Group) ANOVA evaluating response time produced no significant effects, suggesting that response times were similar across time for both groups of participants (see Table 2 for means and standard deviations). Nonparametric Spearman's rank-order correlations were run to evaluate the relationship between accuracy and demographic and cognitive-linguistic variables for the individuals with aphasia. There was no significant correlation between test accuracy and years of education $r_s(9) = .05$, p = .89, months poststroke $r_s(9) = .08, p = .83$, WAB aphasia severity $r_s(9) = .38$, p = .31, CLQT attention, r_s (9) = .57, p = .09, memory r_s (9) = .58, p = .07, executive functions $r_s(9) = .20$, p =

.58, or visuospatial skills $r_s(9) = .45$, p = .18 in this limited sample.

Eye-tracking results

Mean number of observations and proportion log fixation times are reported in Table 3. The linear mixed-effects model evaluating number of observations and the effect of phase, group, and diagnosticity produced a main effect of feature diagnosticity F(3, 210)= 24.88, p < .001 (see Table 4). A significant diagnosticity by phase interaction F(3, 210) = 2.57, p = .05 was produced, with number of observations to features of low diagnosticity decreasing from early to late phases for controls and individuals with aphasia. The group × phase × diagnosticity interaction was nonsignificant, F(3, 210) = 0.09, p = .96.

The beta mixed-effects model evaluating proportion log fixation time produced main effects of group $\beta = -.19$, SE = 0.09, p = .048 and feature with participants showing a

Table 3. Mean number of observations and proportion log fixation times

| | Number of C | Number of Observations | | Fixation Time |
|--------------------|-------------|------------------------|------|---------------|
| | Mean | SD | Mean | SD |
| Controls | | | | |
| Early | | | | |
| HighHigh | 0.75 | 0.43 | 0.30 | 0.04 |
| HighLow | 0.17 | 0.22 | 0.13 | 0.08 |
| Mid | 0.81 | 0.45 | 0.28 | 0.06 |
| Low | 0.73 | 0.41 | 0.29 | 0.07 |
| Late | | | | |
| HighHigh | 0.62 | 0.51 | 0.26 | 0.09 |
| HighLow | 0.28 | 0.37 | 0.15 | 0.13 |
| Mid | 0.79 | 0.45 | 0.30 | 0.10 |
| Low | 0.45 | 0.35 | 0.29 | 0.07 |
| Individuals with a | aphasia | | | |
| Early | | | | |
| HighHigh | 0.64 | 0.47 | 0.29 | 0.08 |
| HighLow | 0.18 | 0.18 | 0.12 | 0.10 |
| Mid | 0.80 | 0.43 | 0.35 | 0.13 |
| Low | 0.50 | 0.33 | 0.23 | 0.09 |
| Late | | | | |
| HighHigh | 0.53 | 0.46 | 0.30 | 0.23 |
| HighLow | 0.31 | 0.37 | 0.19 | 0.14 |
| Mid | 0.73 | 0.67 | 0.27 | 0.24 |
| Low | 0.30 | 0.29 | 0.23 | 0.19 |

| | Number of Observations | | | Proportion Log Fixation Time | | |
|----------------------------|------------------------|-------------------|-------|---------------------------------|-------------------|-------|
| | Estimate | Standard Error | Þ | Estimate | Standard Error | Þ |
| GroupPWA | -0.08 | 0.09 | .38 | - 0.19 | 0.10 | .05 |
| DiagClassHighLow | -0.54 | 0.09 | <.001 | -1.03 | 0.19 | <.001 |
| DiagClassLow | -0.07 | 0.09 | .44 | -0.15 | 0.17 | .36 |
| DiagClassMid | 0.10 | 0.09 | .28 | 0.02 | 0.17 | .89 |
| PhaseLate | -0.13 | 0.09 | .17 | -0.26 | 0.17 | .13 |
| DiagClassHighLow:PhaseLate | 0.24 | 0.13 | .05 | 0.27 | 0.26 | .31 |
| DiagClassLow:PhaseLate | -0.12 | 0.13 | .35 | 0.15 | 0.24 | .53 |
| DiagClassMid:PhaseLate | 0.09 | 0.13 | .49 | -0.01 | 0.24 | .95 |

Table 4. Summary results of mixed-effects models for number of observations and proportion log fixation time

significantly larger proportion log fixation to the feature of HighHigh Diagnosticity relative to HighLow diagnosticity $\beta = -1.03$, *SE* = 0.18, *p* < .001. There were no significant interactions (see Table 4 for coefficients and standard errors).

DISCUSSION

The purpose of this study was to examine patterns of attentional allocation in individuals with aphasia and age-matched controls during a nonlinguistic learning task using eye tracking. This was a replication of an earlier study conducted in young adults (Rehder & Hoffman, 2005); therefore, similar eye-tracking variables were evaluated.

Behavioral analyses revealed that overall accuracy rates of individuals with aphasia and control participants were similar and above chance. Both groups showed increases in performance over the course of learning and carried this over into testing phases. No participants reached criterion of two consecutive blocks without error, however, which suggests weaker overall learning performance compared with the original study by Rehder and Hoffman (2005) in which nearly half of study participants reached criterion. Findings of reduced learning are consistent with prior studies that have identified age-related deficits in category learning (Ashby et al., 2003; Bharani et al., 2016; Maddox, Pacheco, Reeves, Zhu, & Schnyer, 2010; Racine, Barch, Braver, & Noelle, 2006; Ridderinkhof, Span, & Van Der Molen, 2002). Relative to young adults, older adults often produce lower accuracies, longer response times, and slower or reduced development of optimal strategies to support learning. Studies have suggested that reduced learning may be related to reasoning, working memory, or inhibitory control deficits (Ashby, Alfonso-Reese, & Waldron, 1998; Bharani et al., 2016; Maddox et al., 2010; Racine et al., 2006). Studies that have begun to examine nonlinguistic learning in individuals with aphasia sometimes find the learning of individuals with aphasia to be reduced relative to controls (Christiansen et al., 2010; Schuchard & Thompson, 2014; Vallila-Rohter & Kiran, 2013), whereas other studies observe comparable learning across individuals with aphasia and age-matched controls (Goschke et al., 2001; Schuchard & Thompson, 2014, 2017) as we observe here. Analyses of individual participant data often reveal subgroups of individuals with aphasia who learn well and others who score within normal age-matched range (Vallila-Rohter & Kiran, 2013; Zimmerer, Cowell, & Varley, 2014). Characteristics of the learning task and of aphasia profile are

likely to influence performance and likely lead to the aforementioned discrepancies.

Examining eye-tracking results, analyses of number of observations, and proportion log fixation time revealed group differences between individuals with aphasia and controls only on number of observations. Few studies have examined eye gaze behaviors in individuals with aphasia, though Thiessen, Beukelman, Ullman, and Longenecker (2014) collected data that confirmed that people with aphasia present with fixation patterns to images of humans engaged in activity in a manner similar to that observed by prior studies in controls. Interestingly, as described in further detail later, results from our sample of older adults did not replicate all findings previously observed by Rehder and Hoffman (2005) in undergraduate-aged students. Therefore, the small sample of higher level individuals with aphasia included in this study did not differ from older control participants. These participants, however, differed from previously published data collected from young adults performing the same task.

A phase by feature interaction demonstrated that fewer looks were made to the feature of low diagnosticity (the tail) over time. This is consistent with Rehder and Hoffman (2005) and suggests an ability to optimize looking behaviors through the course of learning. Rehder and Hoffman (2005) additionally found that young adult participants produced more and longer fixations to features of high diagnosticity, particularly in the end of training. In contrast to Rehder and Hoffman (2005), current study results demonstrated elevated number and duration of fixations to one of the features of high diagnosticity but not to both. Studies have identified age-related deficits in visual scanning (e.g., Madden, 2007; Zanto & Gazzaley, 2014) and switching (see Wasylyshyn, Verhaeghen, & Sliwinski, 2011). It may be that older individuals had difficulty scanning the four stimulus features or that once they selected features to attend to, the process of shifting attentional allocation to features of high diagnosticity was reduced. Studies have also found reduced inhibition in older adults (Glisky, 2007; Hasher & Zacks, 1988; Hedden & Park, 2001; Salthouse & Meinz, 1995; Sweeney, Rosano, Berman, & Luna, 2001; Valeriani, Ranghi, & Giaquinto, 2003). To focus attention, distractors must be inhibited, a process that has been shown to decline with aging and may have played a role in observed results. Number and proportion of fixations to the feature head were high throughout training in our study. Rehder and Hoffman (2005) also saw relatively high fixations to the head (of medium diagnosticity), which they attributed to its increased salience due to the eye. In the study by Rehder and Hoffman (2005), however, participants showed a gradual decrease in attentional allocation to the head feature of time, which we did not observe. The increased salience of the eye may be difficult to inhibit, allowing for a shift of attention to other, more diagnostic, features.

LIMITATIONS

The sample size included in this study limits the conclusions that can be drawn from results. Furthermore, individuals with aphasia were relatively homogeneous with most presenting with anomic aphasia and relatively low severities as determined by aphasia quotient. Future studies should examine attentional allocation and learning behaviors from a wider sample of individuals with aphasia to determine whether behaviors change as a function of aphasia type or severity. In addition, in the future, stimuli could be modified to reduce the salience of the head, given that this feature (of medium diagnosticity) was looked at on many trials in our sample as well as in the sample collected by Rehder and Hoffman (2005). Interestingly, features of high diagnosticity were on top and below the central body and medium and low diagnosticity features were on the horizontal axis. Visual scanning often takes place from left to right in the act of reading, which may have influenced results. Future studies should manipulate the distribution of high diagnosticity features to determine whether position influences gaze behaviors.

CONCLUSION

Clinically, the current findings draw attention to the importance of considering the visual demands and visual complexity of materials in therapy. In this study, eye gaze behaviors of older adults and individuals with aphasia showed some optimized looking, but this was limited in comparison with prior studies examining the same task in young adults. Older adults and individuals with aphasia may have more difficulty inhibiting distractors and/or shifting visual attention to visually meaningful information. Thus, clinicians may consider using simplified materials that do not include distractors or may provide instruction as to the optimal attentional allocation. The current sample included individuals with mild severities of aphasia. Some therapy tools utilized to support communication for individuals with more severe deficits, such as those utilizing alternative and augmentative communication devices, are likely to incorporate complex visual arrays. Further research is needed to understand how individuals with aphasia engage with such platforms and whether there are ways to enhance factors such as attentional allocation.

REFERENCES

- Arbel, Y., Feeley, E., & Xinyi, H. (2019). Evaluating the relationship between feedback processing and attention allocation in category learning using ERP and eye tracking measures. Manuscript submitted for publication.
- Ashby, F. G., Alfonso-Reese, L. A., & Waldron, E. M. (1998). A neuropsychological theory of multiple systems in category learning. *Psychological Review*, 105(3), 442.
- Ashby, F. G., Noble, S., Filoteo, J. V., Waldron, E. M., & Ell, S. W. (2003). Category learning deficits in Parkinson's disease. *Neuropsychology*, 17(1), 115.
- Bates, D., Maechler, M., Bolker, B., & Walker, S. (2015). Fitting linear mixed-effects models Using lme4. *Journal of Statistical Software*, 67(1), 1-48.
- Bharani, K. L., Paller, K. A., Reber, P. J., Weintraub, S., Yanar, J., & Morrison, R. G. (2016). Compensatory processing during rule-based category learning in older adults. *Aging Neuropsychology and Cognition*, 23(3), 304–326.
- Burgio, F., & Basso, A. (1997). Memory and aphasia. Neuropsychologia, 35(6), 759–766.
- Christiansen, M. H., Kelly, M. L., Shillcock, R. C., & Greenfield, K. (2010). Impaired artificial grammar learning in agrammatism. *Cognition*, 116(3), 382–393.
- Erickson, R. J., Goldinger, S. D., & LaPointe, L. L. (1996). Auditory vigilance in aphasic individuals: Detecting nonlinguistic stimuli with full or divided attention. *Brain and Cognition*, 30(2), 244–253.
- Ettlinger, G., & Moffett, A. M. (1970). Learning in dysphasia. *Neuropsychologia*, 8(4), 465–474.
- Fucetola, R., Connor, L. T., Strube, M. J., & Corbetta, M. (2009). Unravelling nonverbal cognitive performance in acquired aphasia. *Aphasiology*, 23(12), 1418– 1426.

- Glass, A. V., Gazzaniga, M. S., & Premack, D. (1973). Artificial language training in global aphasics. *Neuropsychologia*, 11(1), 95-103.
- Glisky, E. L. (2007). Changes in cognitive function in human aging. In D. R. Riddle (Ed.), *Brain aging: Models methods and mechanisms* (pp. 3–21). Boca Raton, FL: CRC Press/Taylor & Francis.
- Glosser, G., & Goodglass, H. (1990). Disorders in executive control functions among aphasic and other braindamaged patients. *Journal of Clinical and Experimental Neuropsychology*, 12(4), 485–501.
- Goschke, T., Friederici, A. D., Kotz, S. A., & Van Kampen, A. (2001). Procedural learning in Broca's aphasia: Dissociation between the implicit acquisition of spatiomotor and phoneme sequences. *Journal of cognitive neuroscience*, 13(3), 370–388.
- Grossman, M., & Carey, S. (1987). Selective word-learning deficits in aphasia. *Brain and Language*, 32(2), 306– 324.
- Gupta, P., Martin, N., Abbs, B., Schwartz, M., & Lipinski, J. (2006). New word learning in aphasic patients: Dissociating phonological and semantic components. *Brain and Language*, 99(1-2), 8–9.
- Hasher, L., & Zacks, R. T. (1988). Working memory, comprehension, and aging: A review and a new view. *Psychology of Learning and Motivation*, 22, 193-225.
- Hedden, T., & Park, D. (2001). Aging and interference in verbal working memory. *Psychology and Aging*, 16(4), 666.
- Helm-Estabrooks, N. (2001). *Cognitive linguistic quick test*, San Antonio, TX: The Psychological Corporation.
- Helm-Estabrooks, N. (2002). Cognition and aphasia: A discussion and a study. *Journal of Communication Dis*orders, 35(2), 171-186.

- Heuer, S., & Hallowell, B. (2015). A novel eye-tracking method to assess attention allocation in individuals with and without aphasia using a dual-task paradigm. *Journal of Communication Disorders*, 55, 15–30.
- Hula, W. D., & McNeil, M. R. (2008). Models of attention and dual-task performance as explanatory constructs in aphasia. *Seminars in Speech and Language*, 29(3), 169–187.
- Hula, W. D., McNeil, M. R., & Sung, J. E. (2007). Is there an impairment of language-specific attentional processing in aphasia? *Brain and Language*, 103(1), 240– 241.
- Hunting-Pompon, R., Kendall, D., & Bacon Moore, A. (2011). Examining attention and cognitive processing in participants with self-reported mild anomia. *Aphasiology*, 25(6-7), 800–812.
- Kelly, H., & Armstrong, L. (2009). New word learning in people with aphasia. *Aphasiology*, 23(12), 1398-1417.
- Kertesz, A. (2006). Western Aphasia Battery-Revised (WAB-R): Examiner's Manual. PsychCorp. San Antonio, TX: Harcort Assessment Incorporation.
- Knowlton, B. J., Ramus, S. J., & Squire, L. R. (1992). Intact artificial grammar learning in amnesia: Dissociation of classification learning and explicit memory for specific instances. *Psychological Science*, 3(3), 172-179.
- Koenig, P., Smith, E. E., Moore, P., Glosser, G., & Grossman, M. (2007). Categorization of novel animals by patients with Alzheimer's disease and corticobasal degeneration. *Neuropsychology*, 21(2), 193.
- Kuznetsova, A., Brockhoff, P. B., & Christensen, R. H. B. (2017). "ImerTest Package: Tests in linear mixed effects models." *Journal of Statistical Software*, 82(13), 1–26.
- Lambon Ralph, M. A., Snell, C., Fillingham, J. K., Conroy, P., & Sage, K. (2010). Predicting the outcome of anomia therapy for people with aphasia post CVA: Both language and cognitive status are key predictors. *Neuropsychological Rehabilitation*, 20(2), 289-305.
- Laures, J. S. (2005). Reaction time and accuracy in individuals with aphasia during auditory vigilance tasks. *Brain and Language*, 95(2), 353-357.
- Laures, J., Odell, K., & Coe, C. (2003). Arousal and auditory vigilance in individuals with aphasia during a linguistic and nonlinguistic task. *Aphasiology*, 17(12), 1133-1152.
- Lesniak, M., Bak, T., Czepiel, W., Seniów, J., & Członkowska, A. (2008). Frequency and prognostic value of cognitive disorders in stroke patients. *Dementia and Geriatric Cognitive Disorders*, 26(4), 356-363.
- Madden, D. J. (2007). Aging and visual attention. *Current Directions in Psychological Science*, *16*(2), 70-74.
- Maddox, W. T., Pacheco, J., Reeves, M., Zhu, B., & Schnyer, D. M. (2010). Rule-based and informationintegration category learning in normal aging. *Neuropsychologia*, 48(10), 2998–3008.

- Marshall, R. S., Basilakos, A., & Love-Myers, K. (2013). Further evidence of auditory extinction in aphasia. *Journal of Speech Language and Hearing Research*, 56(1):236–249.
- Martin, N., & Saffran, E. M. (1999). Effects of word processing and short-term memory deficits on verbal learning: Evidence from aphasia. *International Journal of Psychology*, 34(5-6), 339-346.
- McNeil, M. R., Odell, K., & Tseng, C. H. (1991). Toward the integration of resource allocation into a general theory of aphasia. *Clinical Aphasiology*, 20, 21– 39.
- Murray, L. L. (1999). Review attention and aphasia: Theory, research and clinical implications. *Aphasiology*, 13(2), 91–111.
- Murray, L. (2000). The effects of varying attentional demands on the word retrieval skills of adults with aphasia, right hemisphere brain damage, or no brain damage. *Brain and Language*, 72(1), 40–72.
- Murray, L. L. (2012). Attention and other cognitive deficits in aphasia: Presence and relation to language and communication measures. *American Journal of Speech-Language Pathology*, 21(2), 851-864.
- Murray, L. L., Holland, A. L., & Beeson, P. M. (1997). Auditory processing in individuals with mild aphasia: A study of resource allocation. *Journal of Speech Lan*guage and Hearing Research, 40(4), 792–808.
- Mysiw, W. J., Beegan, J. G., & Gatens, P. F. (1989). Prospective cognitive assessment of stroke patients before inpatient rehabilitation: The relationship of the Neurobehavioral Cognitive Status Examination to functional improvement. *American Journal of Physical Medicine & Rebabilitation*, 68(4), 168-171.
- Nosofsky, R. M., & Zaki, S. R. (2002). Exemplar and prototype models revisited: Response strategies, selective attention, and stimulus generalization. *Journal of Experimental Psychology: Learning Memory and Cognition*, 28(5), 924.
- Peach, R. K., Newhoff, M., & Rubin, S. S. (1993). Attention in aphasia as revealed by event-related potentials: A preliminary investigation. *Clinical Aphasiology*, 21, 323-333.
- Petry, M. C., Crosson, B., Rothi, L. J. G., Bauer, R. M., & Schauer, C. A. (1994). Selective attention and aphasia in adults: Preliminary findings. *Neuropsychologia*, *32*(11), 1397–1408.
- Potagas, C., Kasselimis, D., & Evdokimidis, I. (2011). Short-term and working memory impairments in aphasia. *Neuropsychologia*, 49(10), 2874–2878.
- Purdy, M. (2002). Executive function ability in persons with aphasia. *Aphasiology*, *16*(4-6), 549-557.
- R Core Team. (2014). *R: A language and environment for statistical computing*. Vienna, Austria: R Foundation for Statistical Computing. Retrieved from http://www.R-project.org/
- Racine, C. A., Barch, D. M., Braver, T. S., & Noelle, D. C. (2006). The effect of age on rule-based category

learning. Aging, Neuropsychology and Cognition, 13(3-4), 411-434.

- Rehder, B., & Hoffman, A. B. (2005). Thirty-something categorization results explained: Selective attention, eyetracking, and models of category learning. *Journal* of *Experimental Psychology: Learning Memory and Cognition*, *31*(5), 811.
- Ridderinkhof, K. R., Span, M. M., & Van Der Molen, M. W. (2002). Perseverative behavior and adaptive control in older adults: Performance monitoring, rule induction, and set shifting. *Brain and Cognition*, 49(3), 382– 401.
- Robertson, I. H., Ridgeway, V., Greenfield, E., & Parr, A. (1997). Motor recovery after stroke depends on intact sustained attention: A 2-year follow-up study. *Neuropsychology*, 11(2), 290.
- Robin, D. A., & Rizzo, M. (1989). The effect of focal cerebral lesions on intramodal and cross-modal orienting of attention. *Clinical Aphasiology*, 18(1), 61–74.
- Salthouse, T. A., & Meinz, E. J. (1995). Aging, inhibition, working memory, and speed. *The Journals of Gerontology Series B: Psychological Sciences and Social Sciences*, 50(6), P297–P306.
- Schuchard, J., & Thompson, C. K. (2014). Implicit and explicit learning in individuals with agrammatic aphasia. *Journal of Psycholinguistic Research*, 43(3), 209– 224.
- Schuchard, J., & Thompson, C. K. (2017). Sequential learning in individuals with agrammatic aphasia: Evidence from artificial grammar learning. *Journal of Cognitive Psychology*, 29(5), 521-534.
- Shohamy, D., Myers, C. E., Onlaor, S., & Gluck, M. A. (2004). Role of the basal ganglia in category learning: How do patients with Parkinson's disease learn?. *Behavioral Neuroscience*, *118*(4), 676.
- Silkes, J. P., McNeil, M. R., & Drton, M. (2004). Simulation of aphasic naming performance in non-brain-damaged adults. *Journal of Speech Language and Hearing Researcb*, 47(3), 610–623.
- Skaug, H., Fournier, D., Nielsen, A., Magnusson, A., & Bolker, B. (2011). glmmADMB: Generalized linear mixed models using AD Model Builder.
- Sweeney, J. A., Rosano, C., Berman, R. A., & Luna, B. (2001). Inhibitory control of attention declines more than working memory during normal aging. *Neurobiology of Aging*, 22(1), 39–47.
- Thiessen, A., Beukelman, D., Ullman, C., & Longenecker, M. (2014). Measurement of the visual attention patterns of people with aphasia: A preliminary investigation of two types of human engagement in photographic images. *Augmentative and Alternative Communication*, 30(2), 120–129.

- Tseng, C. H., McNeil, M. R., & Milenkovic, P. (1993). An investigation of attention allocation deficits in aphasia. *Brain and Language*, 45(2), 276–296.
- Tuomiranta, L., Grönholm-Nyman, P., Kohen, F., Rautakoski, P., Laine, M., & Martin, N. (2011). Learning and maintaining new vocabulary in persons with aphasia: Two controlled case studies. *Aphasiology*, 25(9), 1030-1052.
- Tuomiranta, L., Grönroos, A. M., Martin, N., & Laine, M. (2014). Vocabulary acquisition in aphasia: Modality can matter. *Journal of Neurolinguistics*, 32, 42-58.
- Tuomiranta, L., Rautakoski, P., Rinne, J. O., Martin, N., & Laine, M. (2012). Long-term maintenance of novel vocabulary in persons with chronic aphasia. *Aphasi*ology, 26(8), 1053-1073.
- Valeriani, M., Ranghi, F., & Giaquinto, S. (2003). The effects of aging on selective attention to touch: A reduced inhibitory control in elderly subjects?. *International Journal of Psychophysiology*, 49(1), 75-87.
- Vallila-Rohter, S., & Kiran, S. (2013). Non-linguistic learning and aphasia: Evidence from a paired associate and feedback-based task. *Neuropsychologia*, 51(1), 79–90.
- Vallila-Rohter, S., & Kiran, S. (2015). An examination of strategy implementation during abstract nonlinguistic category learning in aphasia. *Journal of Speech Language and Hearing Research*, 58(4), 1195-1209.
- Villard, S. (2017). Potential implications of attention deficits for treatment and recovery in aphasia. *Perspectives of the ASHA Special Interest Groups*, 2(2), 7-14.
- Villard, S., & Kiran, S. (2015). Between-session intraindividual variability in sustained, selective, and integrational non-linguistic attention in aphasia. *Neuropsychologia*, 66, 204–212.
- Villard, S., & Kiran, S. (2017). To what extent does attention underlie language in aphasia?. *Aphasiology*, 31(10), 1226-1245.
- Villard, S., & Kiran, S. (2018). Between-session and withinsession intra-individual variability in attention in aphasia. *Neuropsychologia*, 109, 95-106.
- Wasylyshyn, C., Verhaeghen, P., & Sliwinski, M. J. (2011). Aging and task switching: A meta-analysis. *Psychology* and Aging, 26(1), 15.
- Zanto, T. P., & Gazzaley, A. (2014). Attention and ageing. In K. Nobre, A. C. Nobre, & S. Kastner (Eds.), *The Oxford bandbook of attention* (pp. 927-971). Oxford, United Kingdom: Oxford University Press.
- Zimmerer, V. C., Cowell, P. E., & Varley, R. A. (2014). Artificial grammar learning in individuals with severe aphasia. *Neuropsychologia*, 53, 25–38.